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Rome et al.

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(54) **AUTHENTICATION VIA TYPING CADENCE, GESTURES, AND QR CODES**

(56) **References Cited**

U.S. PATENT DOCUMENTS

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4,621,334 A 11/1986 Garcia
4,805,222 A * 2/1989 Young G06F 21/316
382/115

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5,557,686 A 9/1996 Brown et al.
6,151,593 A 11/2000 Cho et al.
6,161,185 A 12/2000 Guthrie et al.
6,442,692 B1 8/2002 Zilberman et al.

(Continued)

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OTHER PUBLICATIONS

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U.S. Appl. No. 14/052,562, U.S. Pat. No. 9,430,626 filed Oct. 11,
2013, User Authentication Via Known Text Input Cadence.

(Continued)

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Primary Examiner — Kaveh Abrishamkar

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(74) *Attorney, Agent, or Firm* — Schwegman Lundberg &
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(57) **ABSTRACT**

Related U.S. Application Data

(63) Continuation of application No. 15/134,348, filed on
Apr. 20, 2016, now Pat. No. 10,235,507.

Disclosed herein are techniques for authenticating a user via
gestures, QR codes, and passphrases generated to incorpo-
rate typing habits of the user. A passphrase system generates
a one-time use passphrase, which incorporates hallmarks
and/or quirks of the user's typing, and presents the generated
passphrase as an authentication challenge to authenticate as
the user. If metrics collected during the authentication chal-
lenge are statistically similar to metrics of the user's typing,
the authentication succeeds; otherwise, the authentication
fails. A user's gesture habits during input of an authentica-
tion drawing may be used as a target for future authentica-
tion attempts. A user's input motions (typing and/or ges-
tures) may be converted into a secure QR code; a different
host device may use the secure QR code to obtain the target
metrics for future authentication attempts of the user.

(60) Provisional application No. 62/150,213, filed on Apr.
20, 2015.

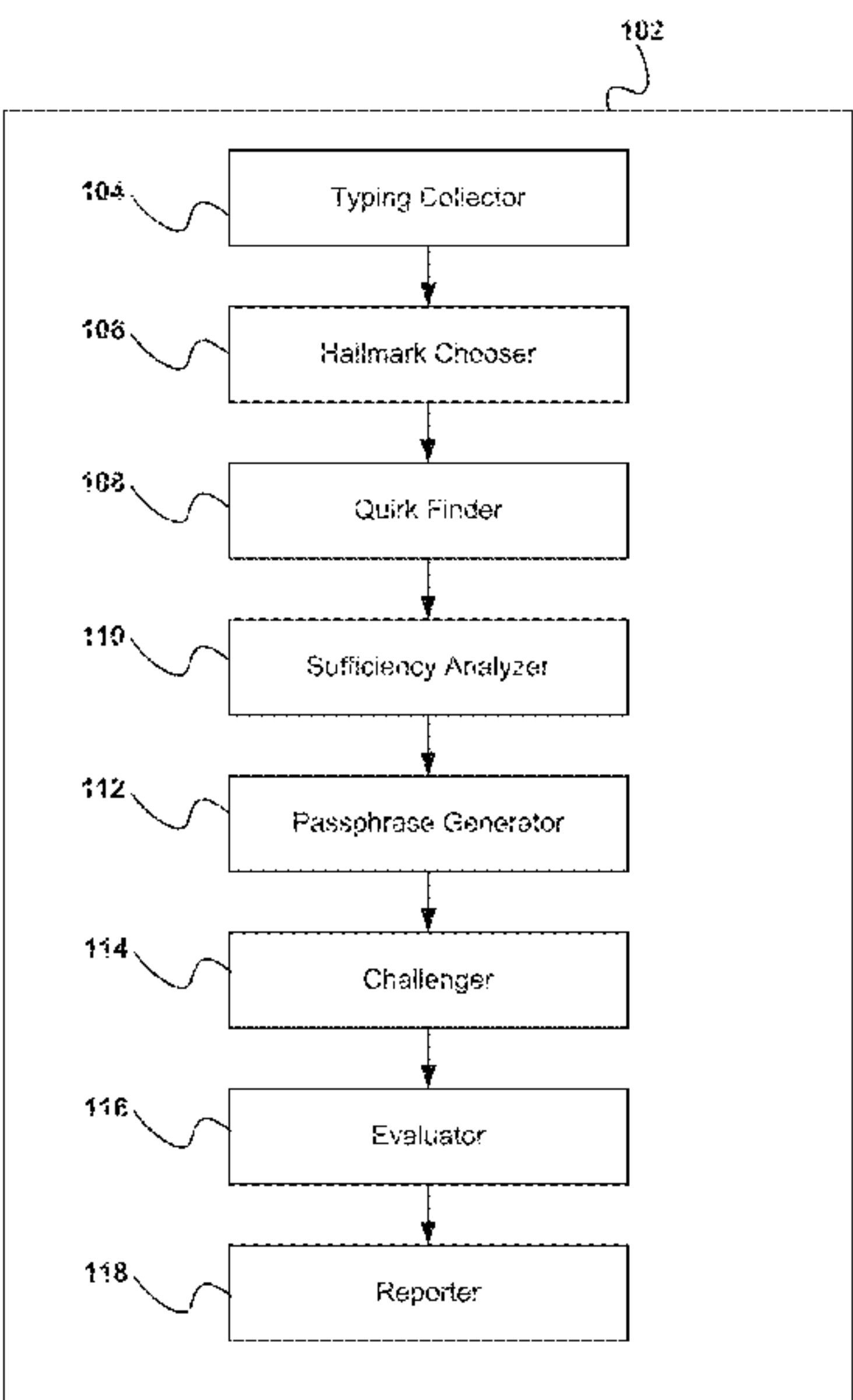
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(52) **U.S. Cl.**
CPC **G06F 21/316** (2013.01); **G06F 21/46**
(2013.01)

(58) **Field of Classification Search**

None
See application file for complete search history.

16 Claims, 18 Drawing Sheets



(56)

References Cited

U.S. PATENT DOCUMENTS

8,141,775	B1 *	3/2012	Aidasani	G06Q 20/3223	235/379
9,355,236	B1 *	5/2016	Kratz	G06F 21/31	
9,430,626	B1 *	8/2016	Rome	G06F 21/32	
9,704,361	B1 *	7/2017	Hazlewood	G08B 21/22	
10,229,255	B2	3/2019	Rome et al.		
10,235,507	B1 *	3/2019	Rome	G06F 21/46	
2003/0228048	A1 *	12/2003	Okada	G06T 7/62	382/144
2004/0034788	A1	2/2004	Ross		
2004/0059950	A1	3/2004	Bender et al.		
2004/0187037	A1	9/2004	Checco		
2006/0153364	A1	7/2006	Beeson		
2006/0274032	A1 *	12/2006	Mao	A63F 13/06	345/156
2006/0280339	A1	12/2006	Cho		
2008/0028231	A1	1/2008	Bender		
2008/0209229	A1	8/2008	Ramakrishnan et al.		
2009/0134972	A1	5/2009	Wu, Jr. et al.		
2009/0262074	A1 *	10/2009	Nasiri	G06F 1/1626	345/158
2012/0051605	A1 *	3/2012	Nagar	G06K 9/00389	382/124
2012/0098750	A1	4/2012	Allen et al.		
2013/0111580	A1 *	5/2013	Checco	G06F 21/32	726/19
2013/0265218	A1 *	10/2013	Moscarillo	G06F 3/04886	345/156
2013/0347099	A1	12/2013	Smith		
2014/0215550	A1 *	7/2014	Adams	G06F 21/316	726/1
2014/0310804	A1 *	10/2014	Apostolos	H04L 63/0861	726/19
2016/0253486	A1	9/2016	Sarkar et al.		
2016/0292410	A1 *	10/2016	Lu	G06F 21/32	
2016/0364558	A1	12/2016	Rome et al.		
2020/0057706	A1	2/2020	Rome et al.		

OTHER PUBLICATIONS

U.S. Appl. No. 15/248,174, U.S. Pat. No. 10,229,255 filed Aug. 26, 2016, User Authentication.

U.S. Appl. No. 16/267,723, filed Feb. 5, 2019, User Authentication.

U.S. Appl. No. 16/544,397, filed Aug. 19, 2019, User Effort Detection.

U.S. Appl. No. 15/134,348, U.S. Pat. No. 10,235,507 filed Apr. 20, 2016, Authentication Via Typing Cadence, Gestures, & QR Codes.

“U.S. Appl. No. 14/052,562, Non Final Office Action dated Dec. 31, 2014”, 22 pgs.

“U.S. Appl. No. 14/052,562, Response filed Jan. 13, 2015 to Non-Final Office Action dated Dec. 31, 2014”, 11 pgs.

“U.S. Appl. No. 14/052,562, Non Final Office Action dated Apr. 6, 2015”, 6 pgs.

“U.S. Appl. No. 14/052,562, Amendment and Response filed Aug. 6, 2015 to Non-Final Office Action dated Apr. 6, 2015”, 22 pgs.

“U.S. Appl. No. 14/052,562, Final Office Action dated Oct. 15, 2015”, 23 pgs.

“U.S. Appl. No. 14/052,562, Response filed Mar. 15, 2016 to Final Office Action dated Oct. 15, 2015”, 13 pgs.

“U.S. Appl. No. 14/052,562, Notice of Allowance dated May 24, 2016”, 10 pgs.

“U.S. Appl. No. 15/248,174, Non Final Office Action dated Apr. 6, 2017”, 13 pgs.

“U.S. Appl. No. 15/248,174, Response filed Oct. 5, 2017 to Non Final Office Action dated Apr. 6, 2017”, 13 pgs.

“U.S. Appl. No. 15/248,174, Final Office Action dated Nov. 21, 2017”, 19 pgs.

“U.S. Appl. No. 15/134,348, Non Final Office Action dated Dec. 4, 2017”.

“U.S. Appl. No. 15/248,174, Response filed Jan. 26, 2018 to Final Office Action dated Nov. 21, 2017”, 17 pgs.

“U.S. Appl. No. 15/134,348, Response filed Apr. 4, 2018 to Non Final Office Action dated Dec. 4, 2017”, 13 pgs.

“U.S. Appl. No. 15/248,174, Non Final Office Action dated May 14, 2018”, 16 pgs.

“U.S. Appl. No. 15/134,348, Final Office Action dated Jul. 16, 2018”, 19 pgs.

“U.S. Appl. No. 15/248,174, Response filed Aug. 14, 2018 to Non Final Office Action dated May 14, 2018”, 16 pgs.

“U.S. Appl. No. 15/134,348, Response filed Sep. 17, 2018 to Final Office Action dated Jul. 16, 2018”, 14 pgs.

“U.S. Appl. No. 15/134,348, Advisory Action dated Sep. 28, 2018”, 3 pgs.

“U.S. Appl. No. 15/248,174, Notice of Allowance dated Oct. 17, 2018”, 9 pgs.

“U.S. Appl. No. 15/134,348, Notice of Allowance dated Oct. 30, 2018”, 5 pages.

Chok, Nian Shong, “Pearson’s Versus Spearman’s and Kendall’s Correlation Coefficients for Continuous Data”, Masters Thesis, University of Pittsburgh, (2010), 53 pgs.

Ghali, S, “Introduction to Geometric Computing”, Chapter 13, (2008), 143-147.

Joyce, R., “Identity Authentication Based on Keystroke Latencies”, Communications of the ACME, vol. 33, No. 2, (Feb. 1990), 168-176.

Kang, P, “Continual Retraining of Keystroke Dynamics Based Authenticator”, In: Lee SW., Li S.Z. (eds) Advances in Biometrics. ICB 2007. Lecture Notes in Computer Science, vol. 4642. Springer, Berlin, Heidelberg., (2007), 9 pgs.

Mandujano, S, “Deterring Password Sharing: User Authentication via Fuzzy c-Means Clustering Applied to Keystroke Biometric Data.”, Proceedings of the Fifth Mexican International Conference in Computer Science (ENC’04)., (2004), 7 pgs.

Monrose, F., “Keystroke dynamics as a biometric for authentication”, Future Generation Computer Systems, 16, (2000), 351-359.

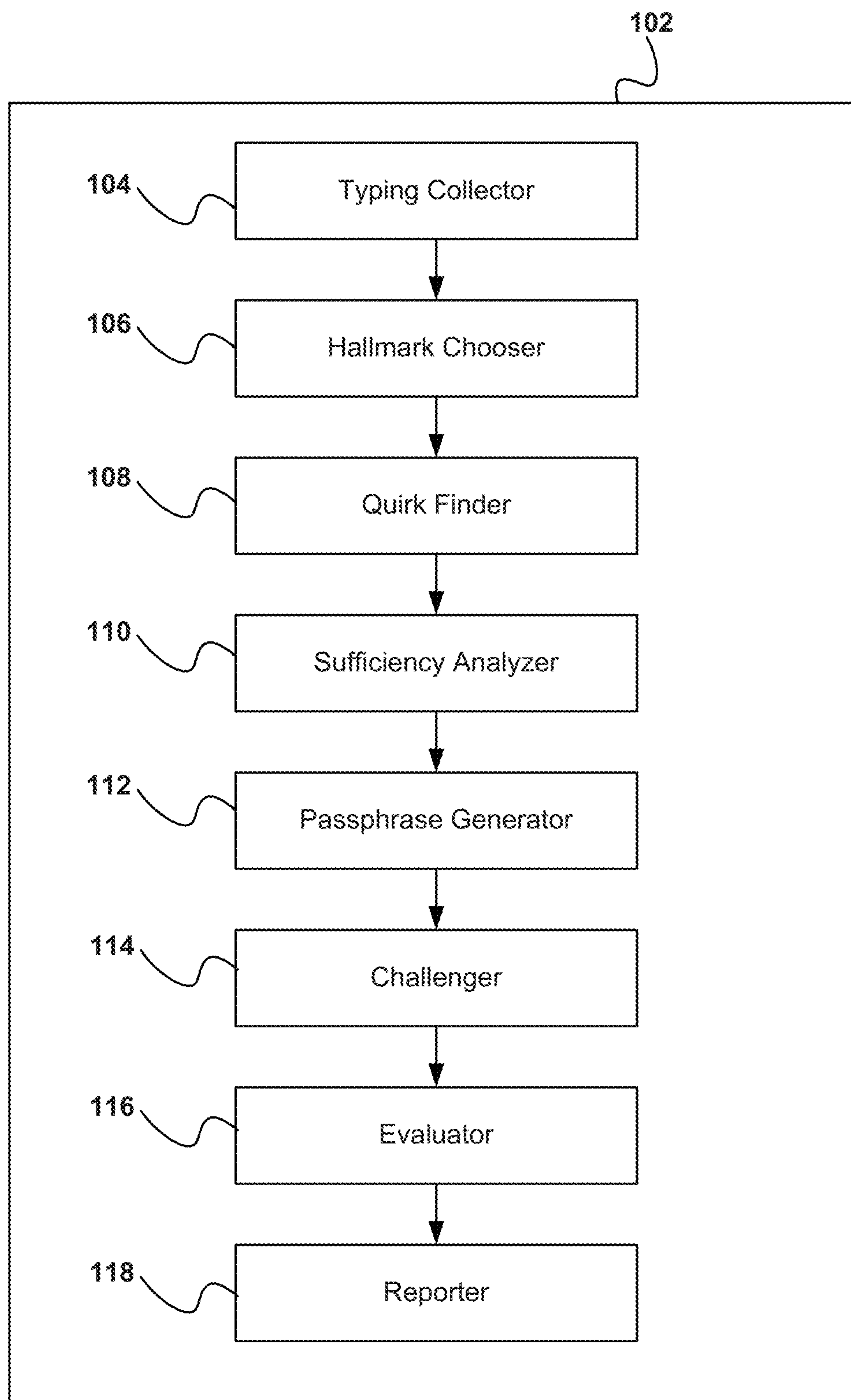
Weisstein, Eric W, “Barycentric Coordinates”, From MathWorld-A Wolfram Web Resource, Online Retrieved from the internet: <http://mathworld.wolfram.com/BarycentricCoordinates.html>, (Oct. 9, 2018).

Wikidiff, “Centroid vs Barycenter—What’s the difference”, [Online] Retrieved from the internet: <https://wikidiff.com/barycenter/centroid>, (Oct. 9, 2018).

Wikipedia, “Centroid”, Wikipedia contributors, The Free Encyclopedia, [Online] Retrieved from the internet: <https://en.wikipedia.org/w/index.php?title=Centroid&oldid=862297606>, (Oct. 9, 2018).

Wikipedia, “Barycenter”, Wikipedia contributors, The Free Encyclopedia, [Online] Retrieved from the internet: <https://en.wikipedia.org/w/index.php?title=Barycenter&oldid=860897006>, (Oct. 9, 2018).

* cited by examiner

**FIG. 1A**

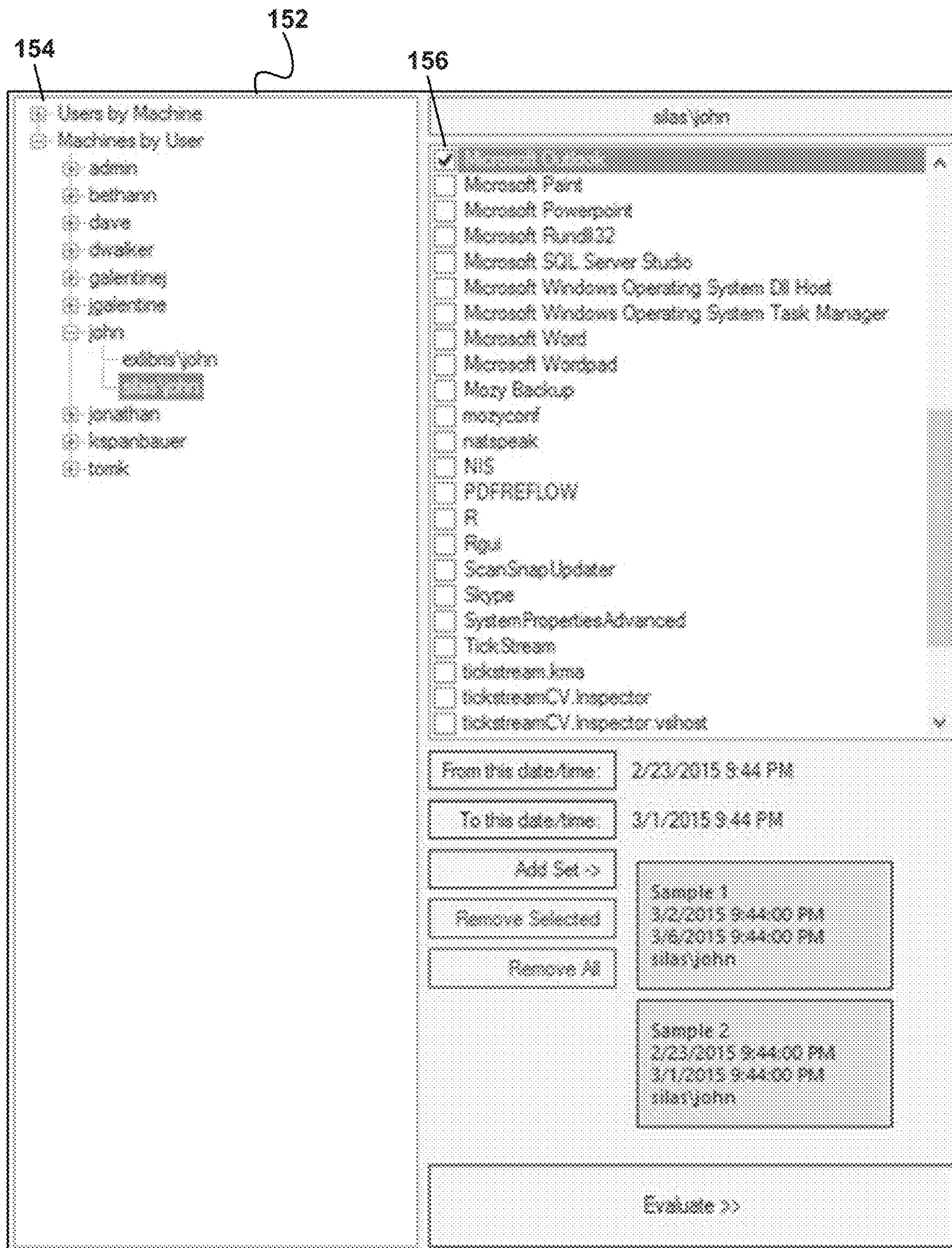


FIG. 1B

200

Session					
Alignment					
Composite					
END					
Statistics					
Metrics					
Session					
Value					
Total Pairs					
172					
Total Qualified Pairs					
112					
Matching Qualified Pairs					
77					
Ratio of Matching to Total Pairs					
45%					
Comparison Ratio (Sample 1/Sample 2)					
98%					
Comparison Observations (Sample 1/Sample 2)					
77					
Qualified Metrics in Comparison					
40%					
Bilateral Overlap					
96%					
Sample 2 Characteristics Passed					
137					
Sample 1 Characteristics Passed					
112					
Sample 1/Sample 2 Amplitude Result					
6					
Sample 1/Sample 2 avgCor					
94.75 %					
Sample 1/Sample 2 avgCor					
94.75 %					
Sample 1/Sample 2 avgEndCor					
82.38 %					
Sample 1/Sample 2 avgEndCor					
82.38 %					
Sample 1/Sample 2 Calculations					
Sample 1/Sample 2 Confidence					
97%					
Sample 1/Sample 2 Consistency					
112					
Sample 1/Sample 2 First Amplitude					
3038					
Sample 1/Sample 2 First Pair Average Of Coeffs					
38.7 %					
Sample 1/Sample 2 First Pair Weighted CoefVar Total					
40.0 %					
Sample 1/Sample 2 First Qual					
112					
Sample 1/Sample 2 First Speed					
13841					
Sample 1/Sample 2 First Yield Average Of Ranks					
6					
Sample 1/Sample 2 First SecondDifferenceVector					
332.7 %					
Sample 1/Sample 2 Null Hypothesis					
11.5 %					
Sample 1/Sample 2 Overlap Match Ratio					
81.8 %					
Sample 1/Sample 2 Projection					
96%					
Sample 1/Sample 2 Ratio Matching to Total					
44.8 %					
Sample 1/Sample 2 Ratio Qualified					
72%					
Sample 1/Sample 2 Second Amplitude					
3208					
Sample 1/Sample 2 Second Pair Average Of Coeffs					
32.7 %					
Sample 1/Sample 2 Second Pair Weighted CoefVar Total					
34.7 %					
Sample 1/Sample 2 Second Qual					
137					
Sample 1/Sample 2 Second Speed					
17882					
Sample 1/Sample 2 Second Yield Average Of Ranks					
5.3					
Sample 1/Sample 2 Session Size Ratio					
81.8 %					
Sample 1/Sample 2 Speed Result					
2.9					
Sample 1/Sample 2 Total Session Metrics					
172					
Sample 1/Sample 2 Weighted Match Ratio					
41.5 %					
Sample 1/Sample 2 Weighted Null Hypothesis					
107%					
Sample 1/Sample 2 yCor					
94.48 %					
Sample 1/Sample 2 yCor					
94.48 %					
Sample 1/Sample 2 yEndCor					
78.97 %					
Sample 1/Sample 2 yEndCor					
78.97 %					
Sample 1/Sample 2 xCor					
95.04 %					
Sample 1/Sample 2 xCor					
95.04 %					

202

FIG. 2

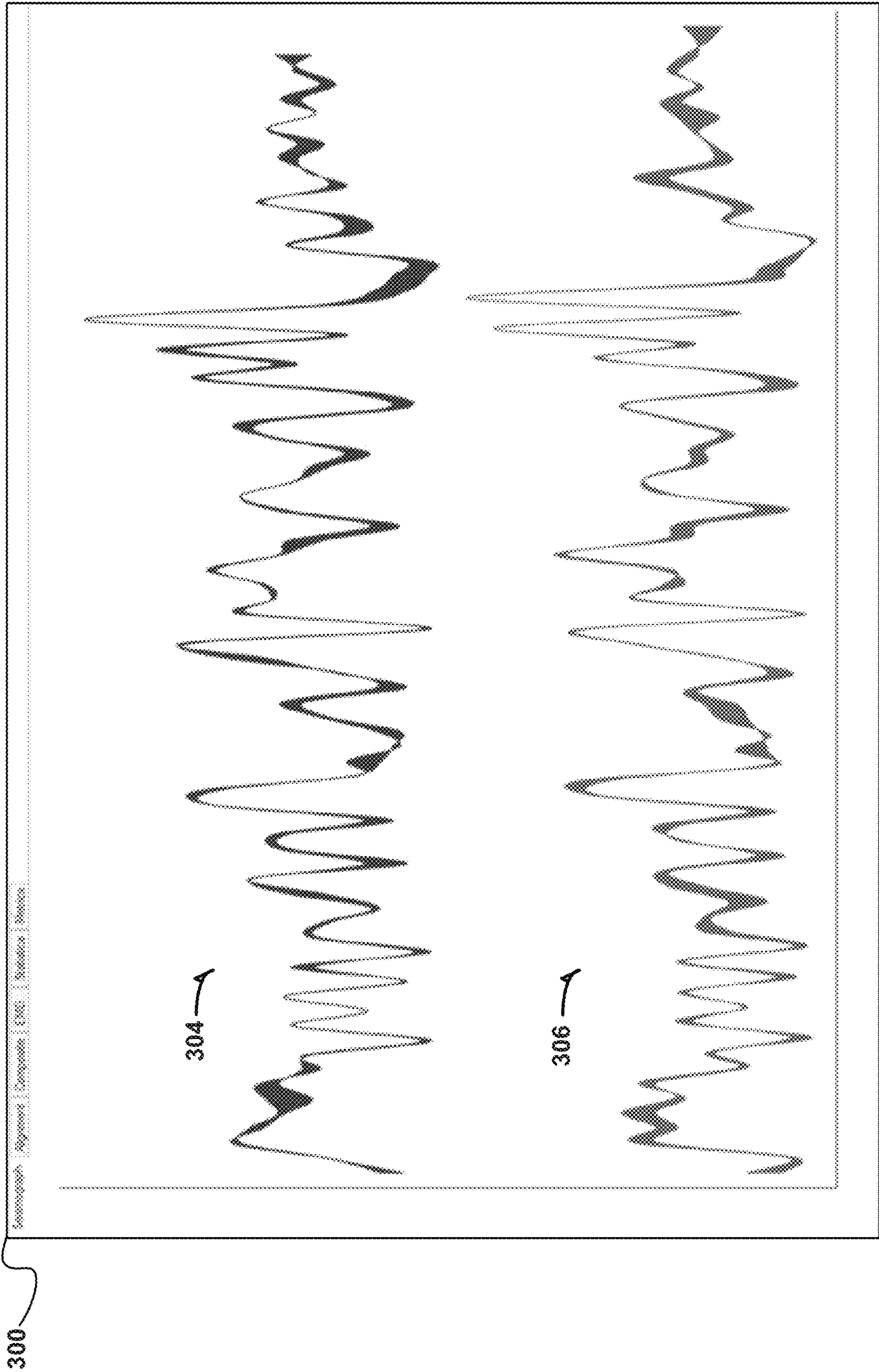


FIG. 3

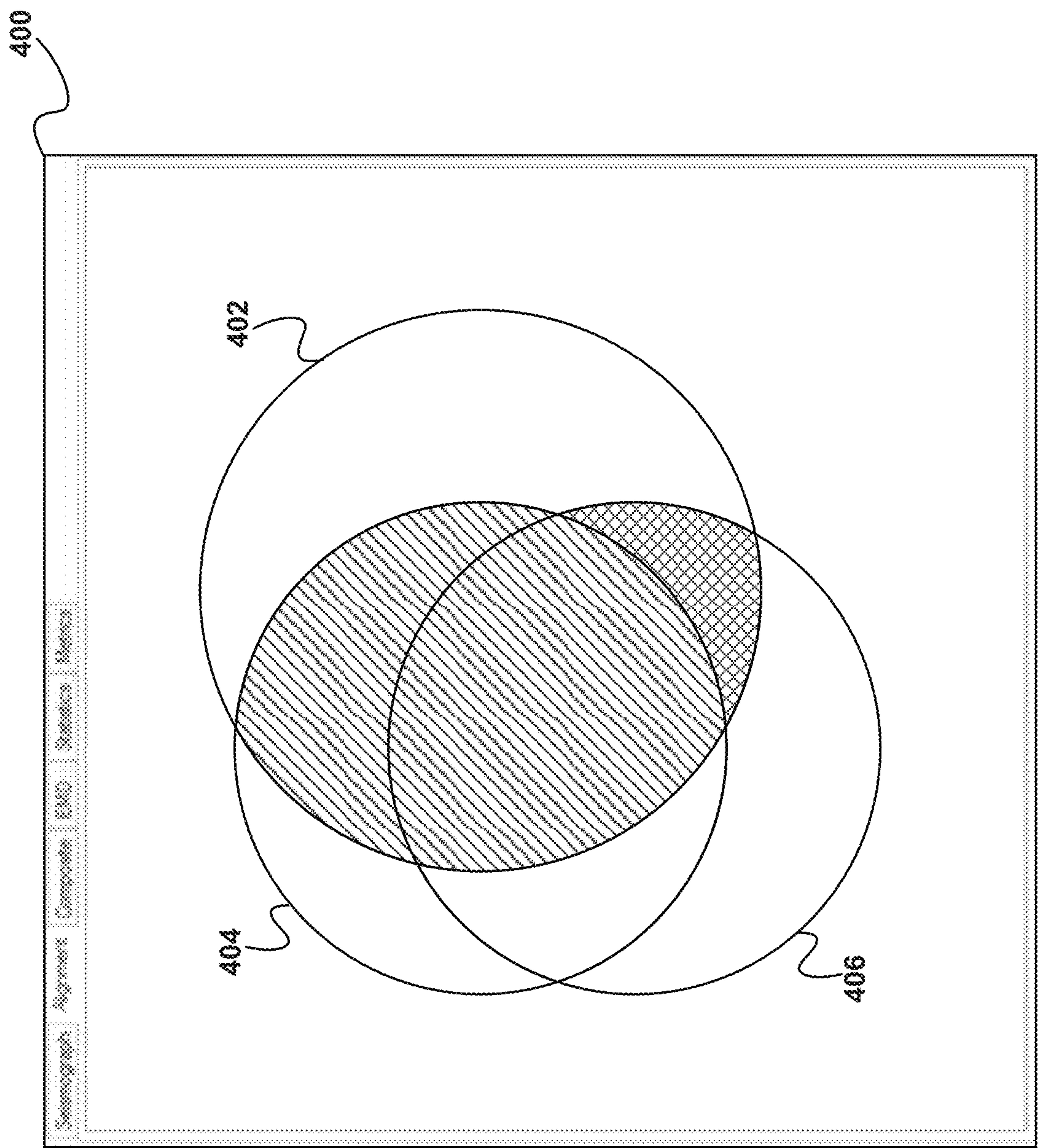


FIG. 4

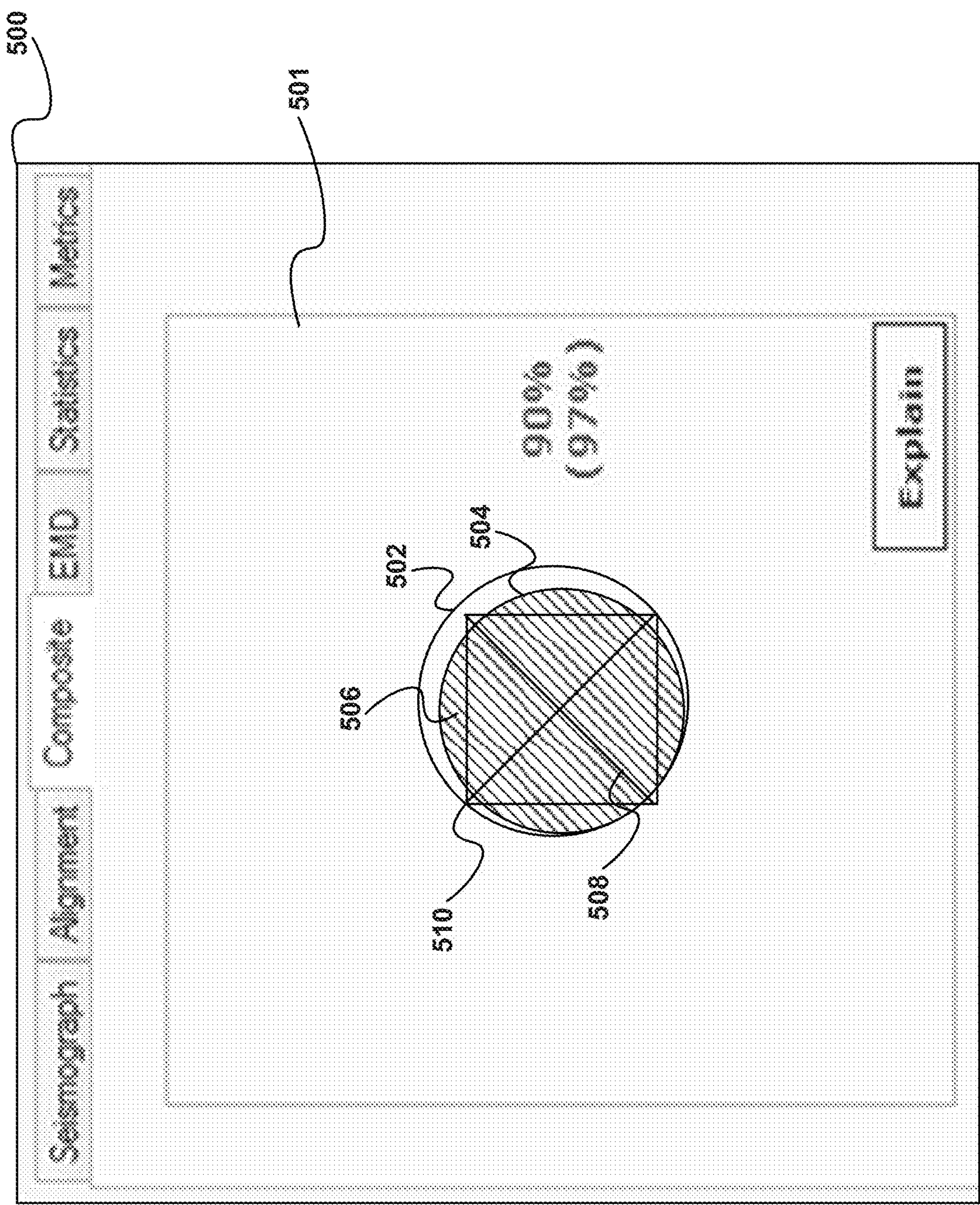


FIG. 5

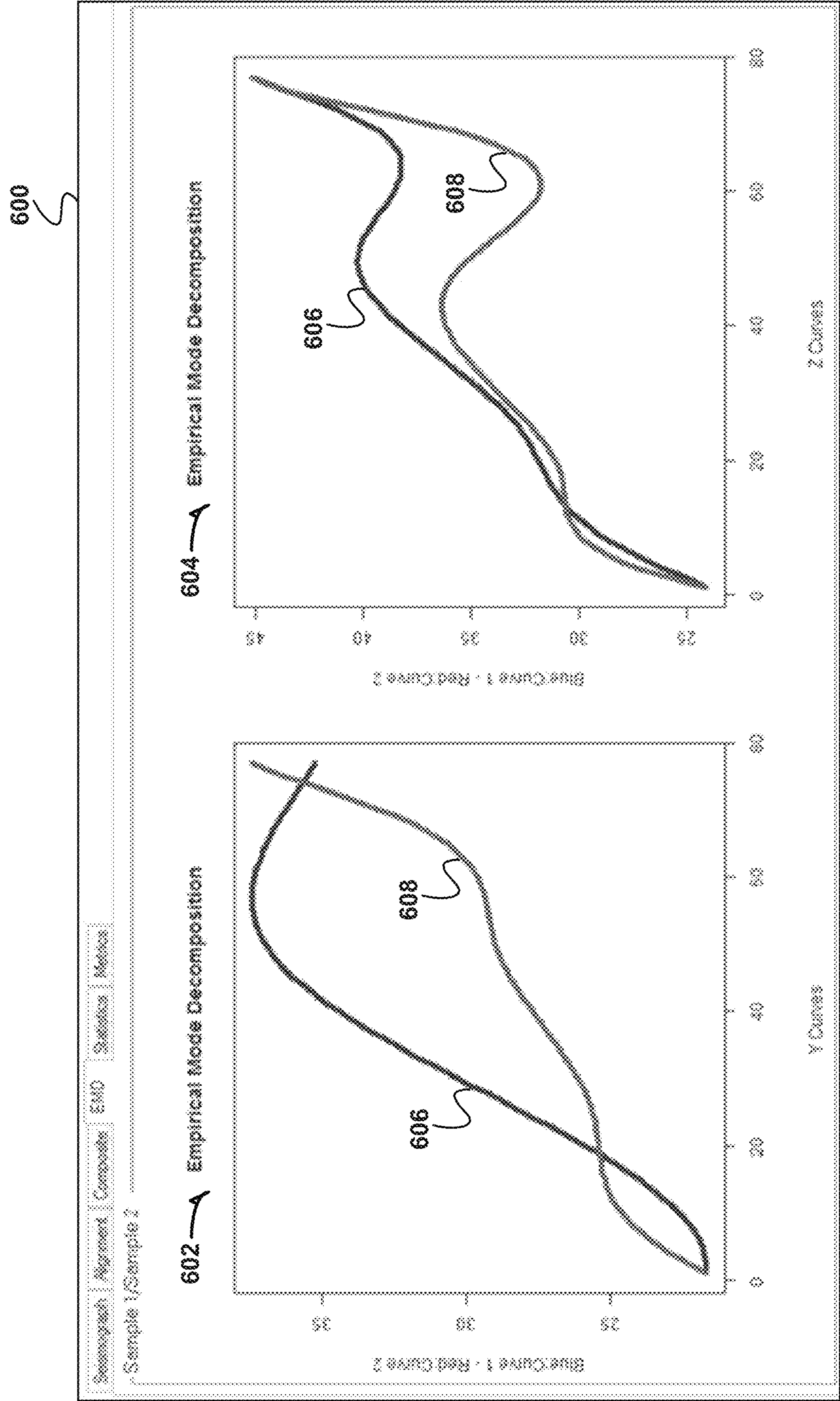


FIG. 6

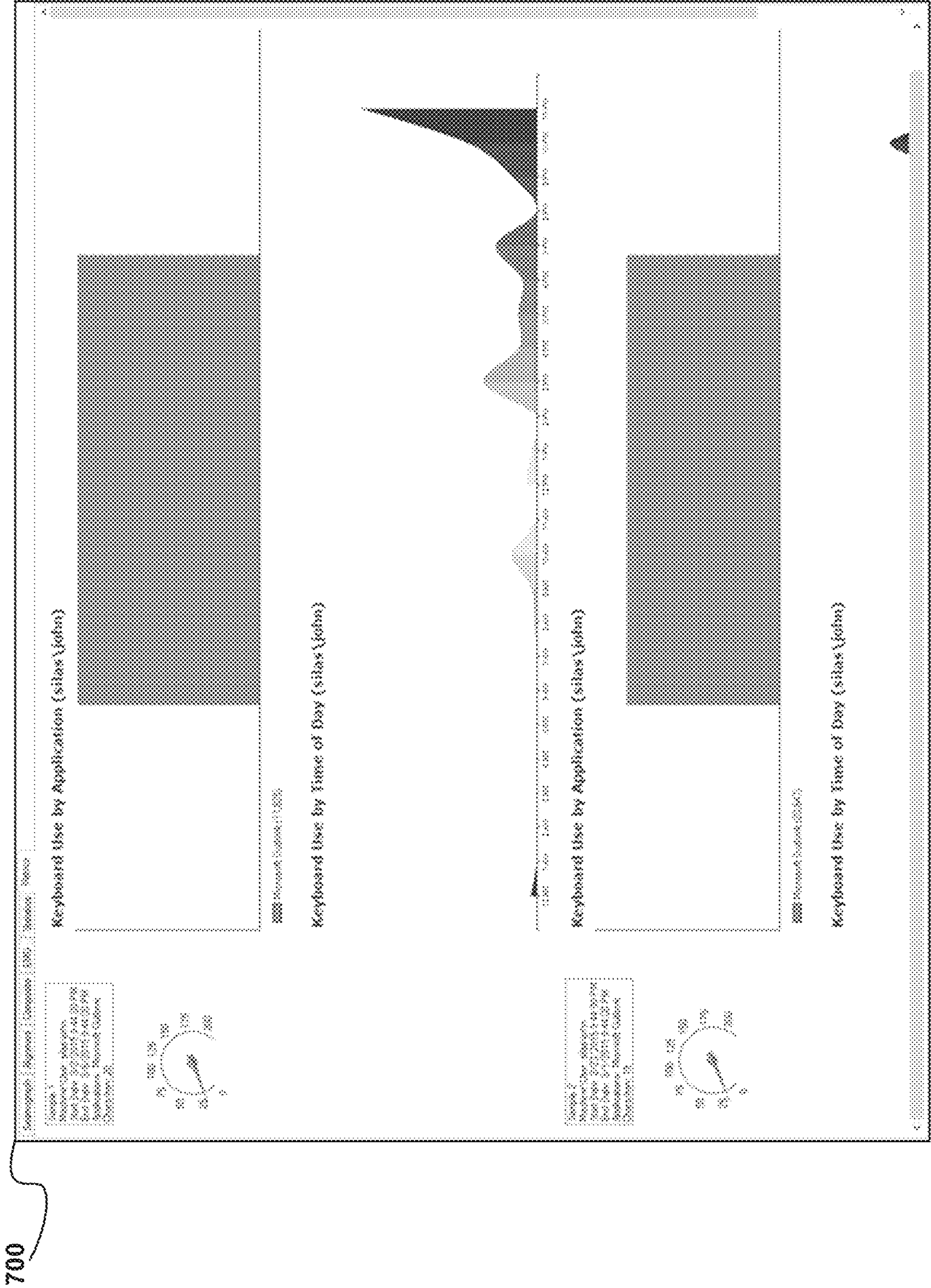
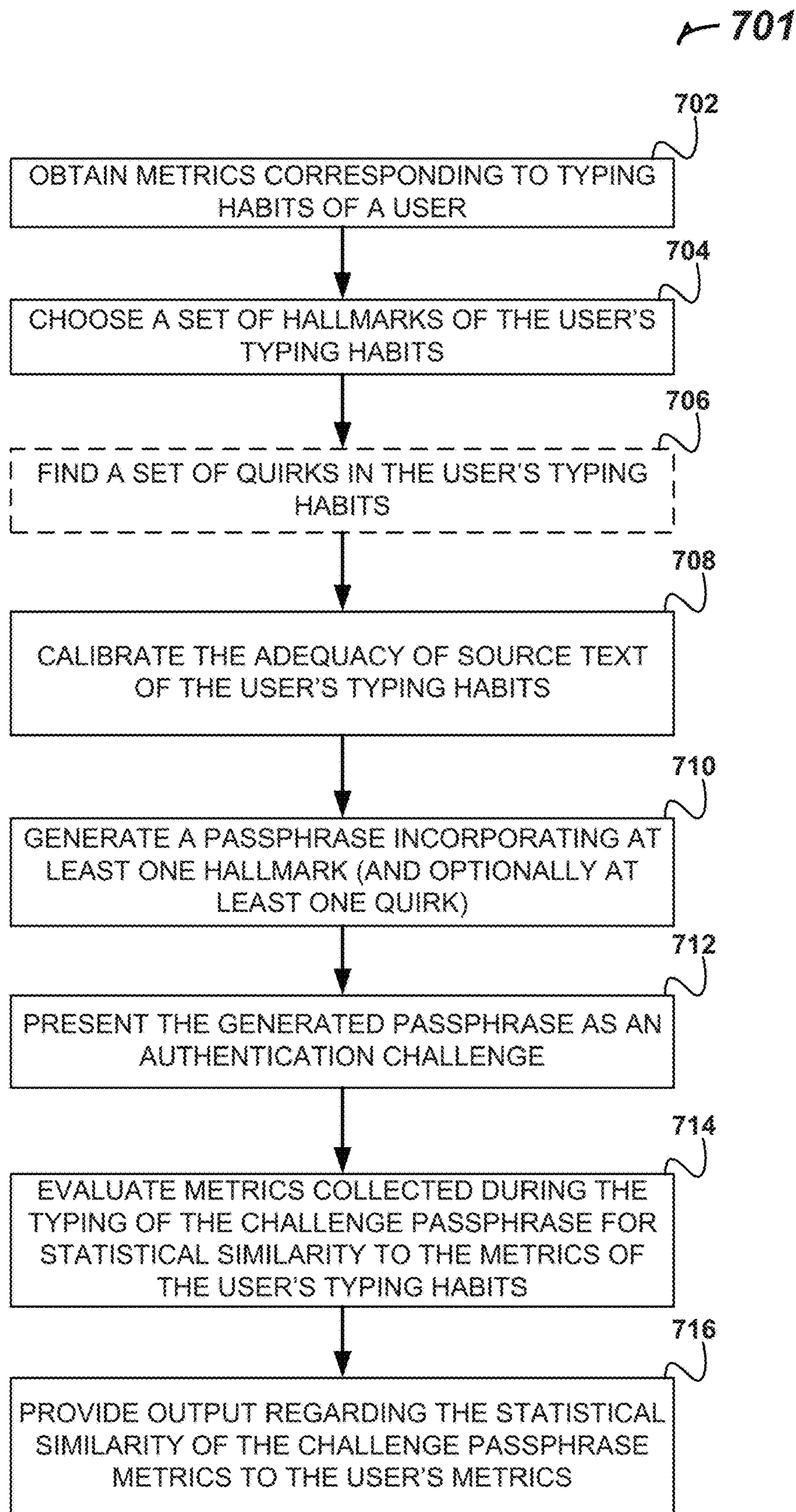


FIG. 7A

**FIG. 7B**

800

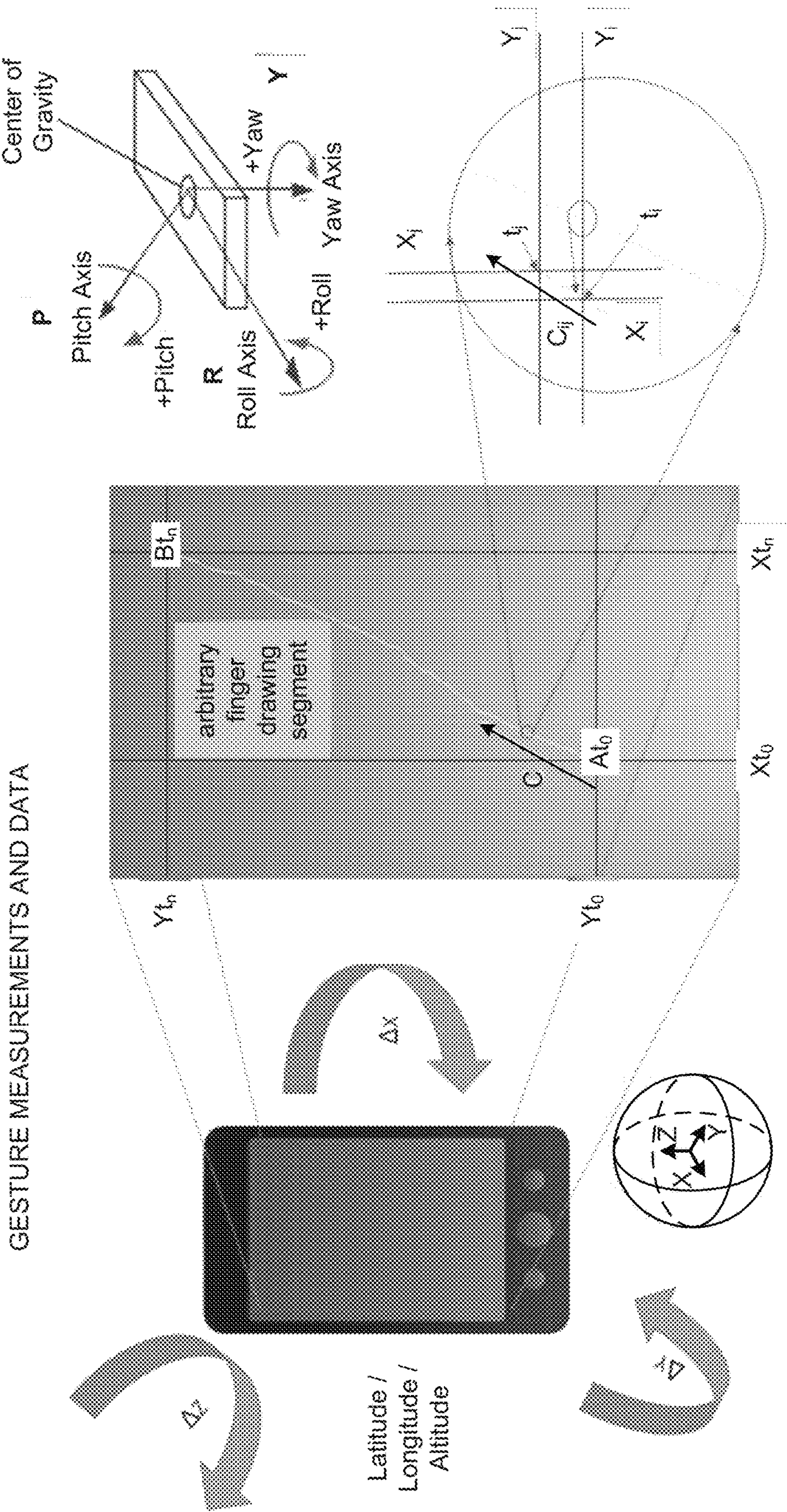


FIG. 8

900

GESTURE MEASUREMENTS AND DATA

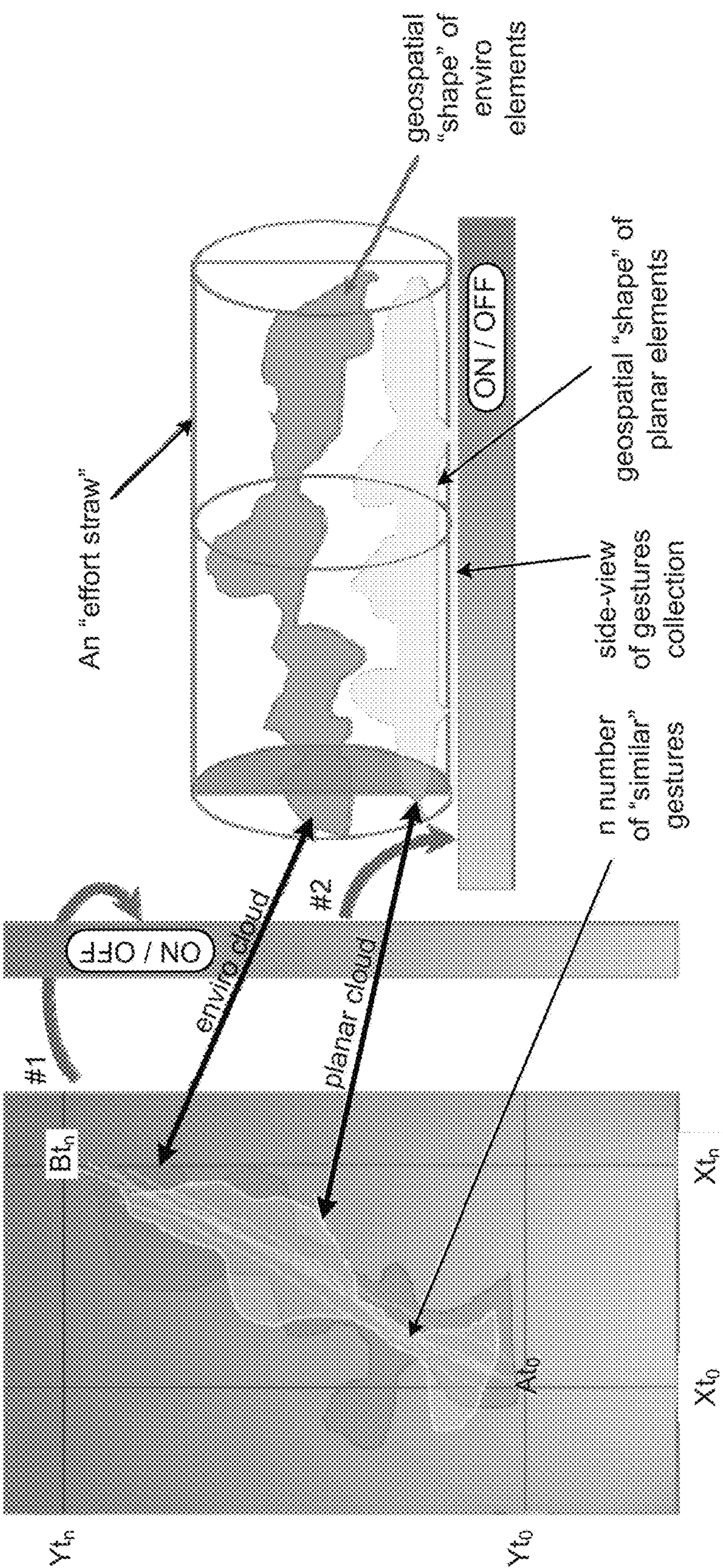


FIG. 9

1000

GESTURE COMPOSITION

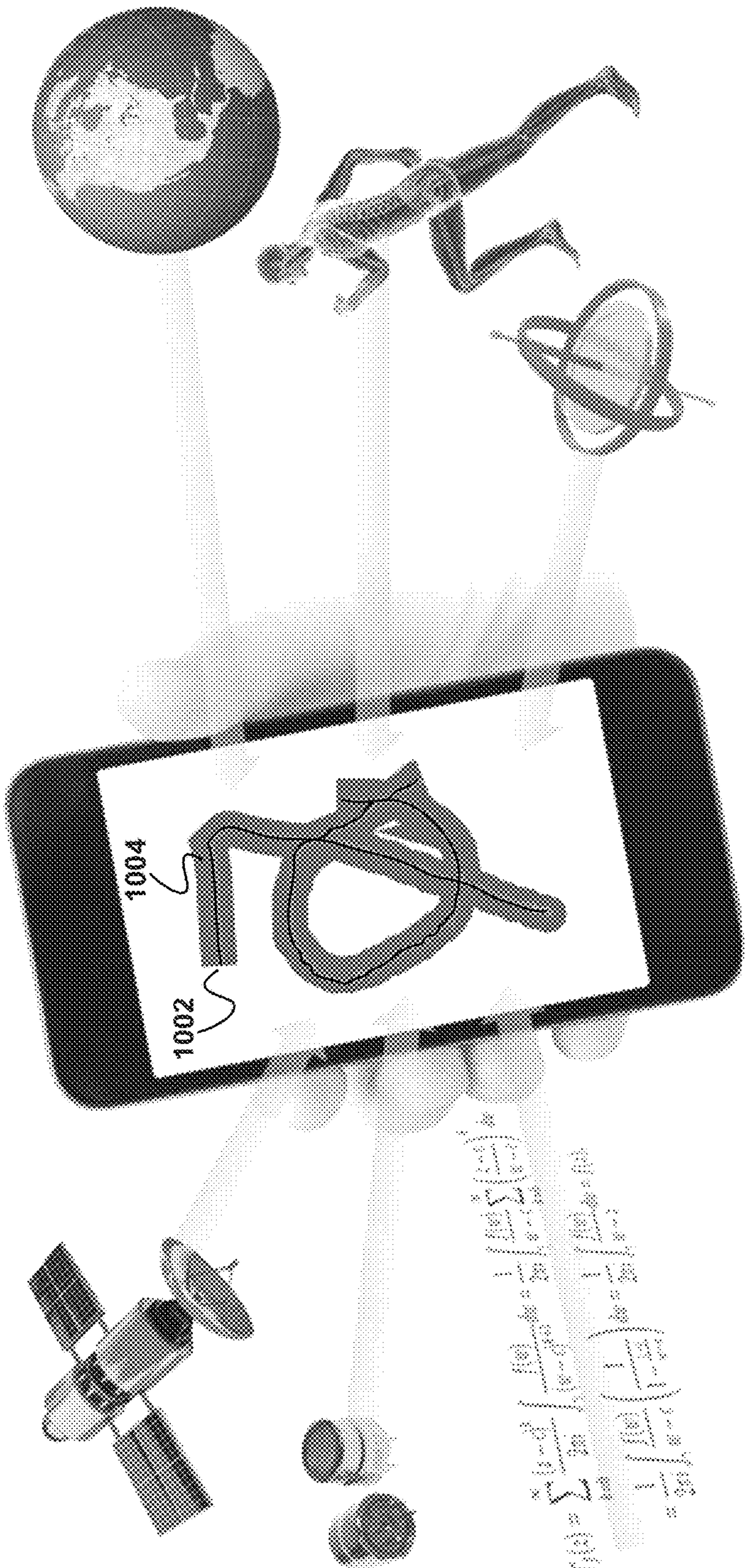


FIG. 10

$$\begin{aligned} T(x) &= \sum_{k=0}^{\infty} \frac{(x-d)^k}{k!} \int_0^{\infty} \frac{f(t)}{(x-d)^{k+1}} dt = \frac{1}{k!} \int_0^{\infty} \frac{f(t)}{(x-d)^{k+1}} dt \\ &= \frac{1}{k!} \int_0^{\infty} \frac{f(t)}{(x-d)^{k+1}} dt = \frac{1}{k!} \int_0^{\infty} \frac{f(t)}{(x-d)^{k+1}} dt \end{aligned}$$

1-1100

MATCHMAKING

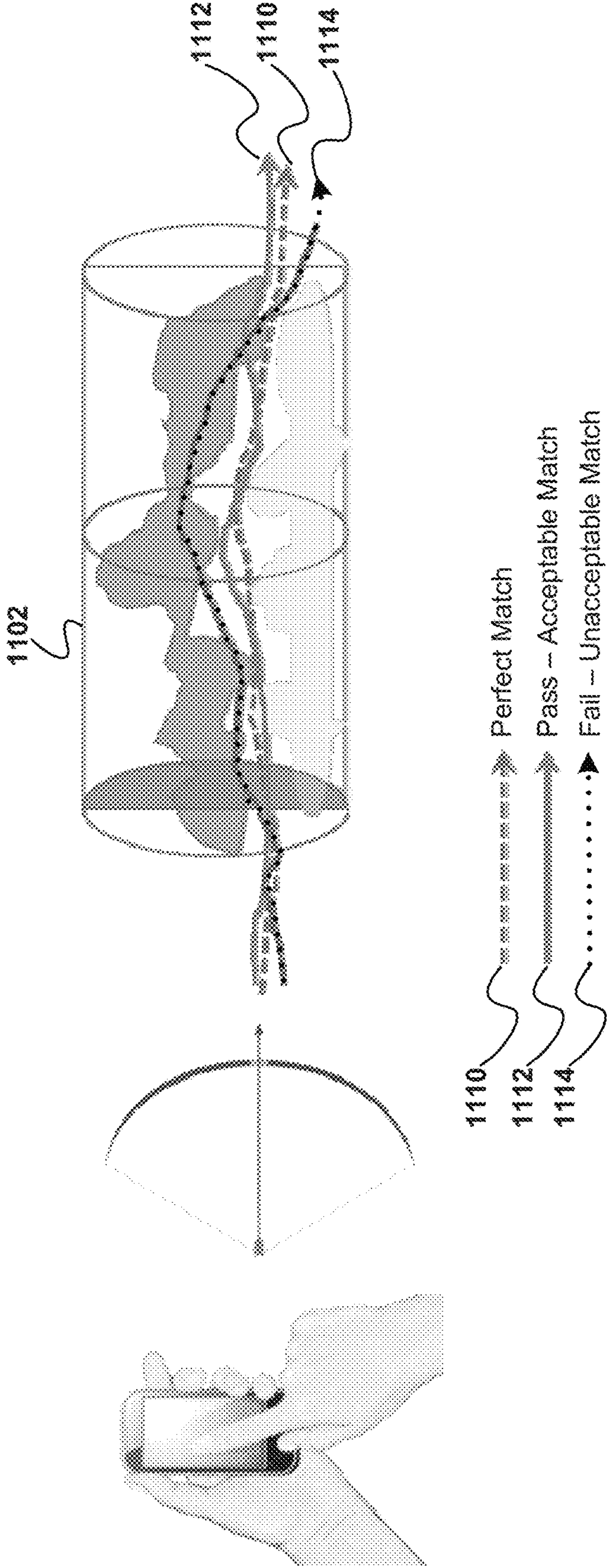


FIG. 11

1200

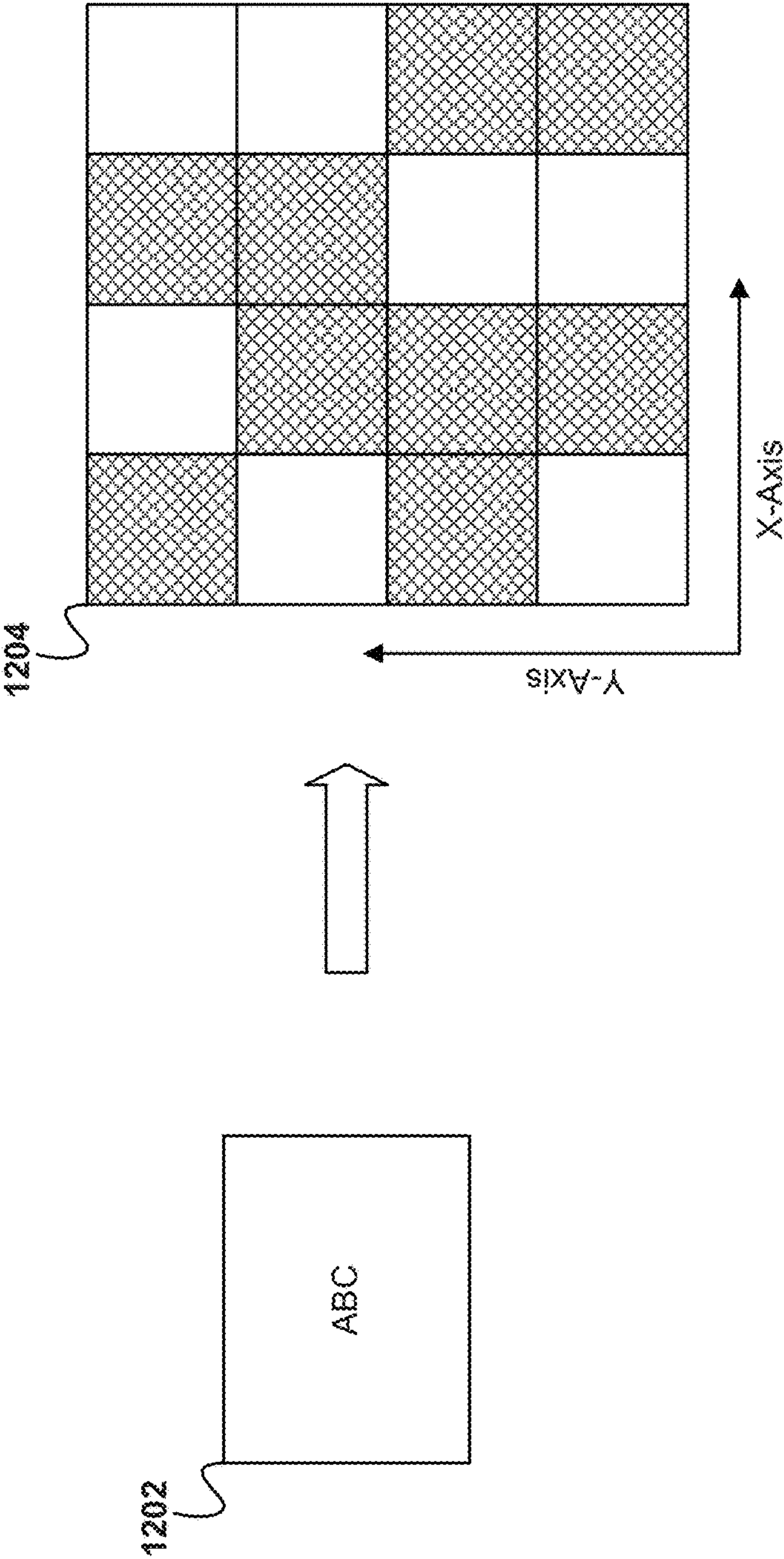


FIG. 12

1300

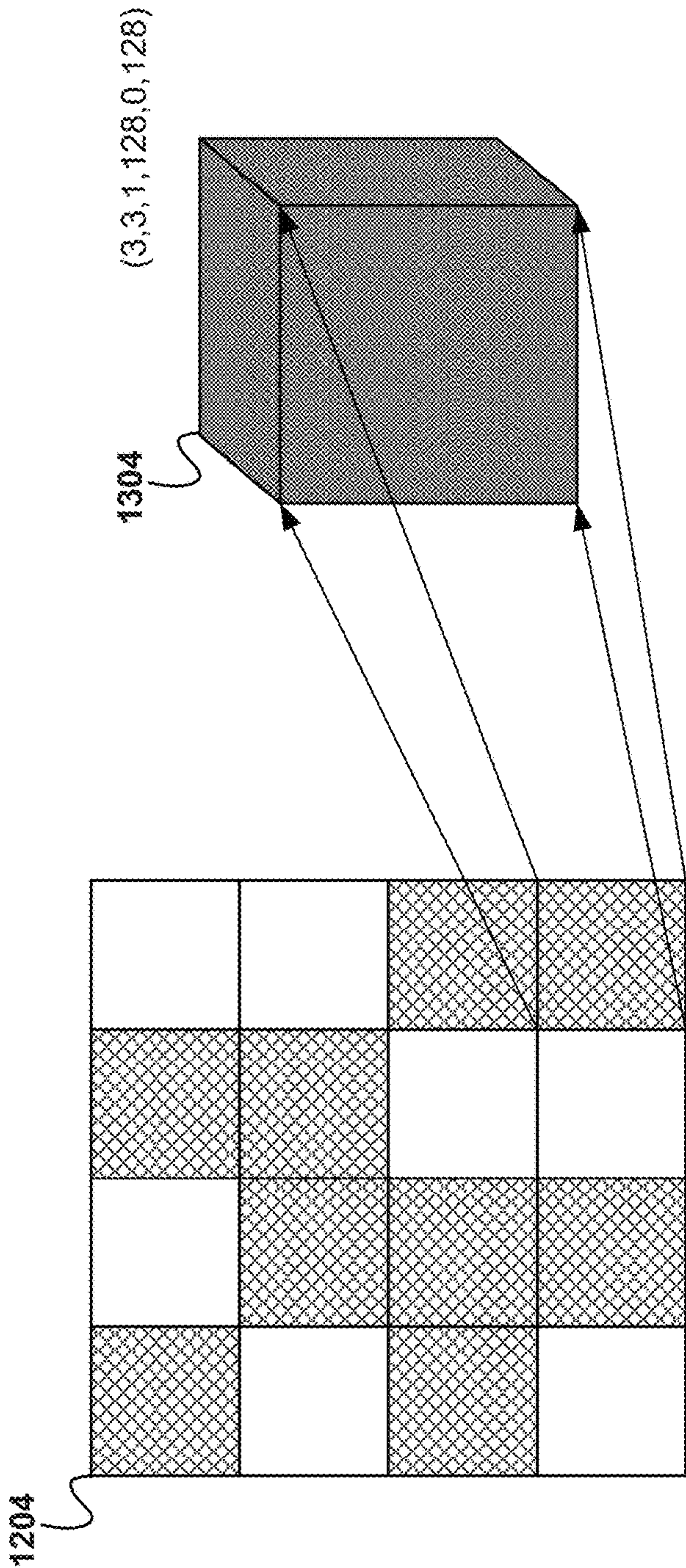


FIG. 13

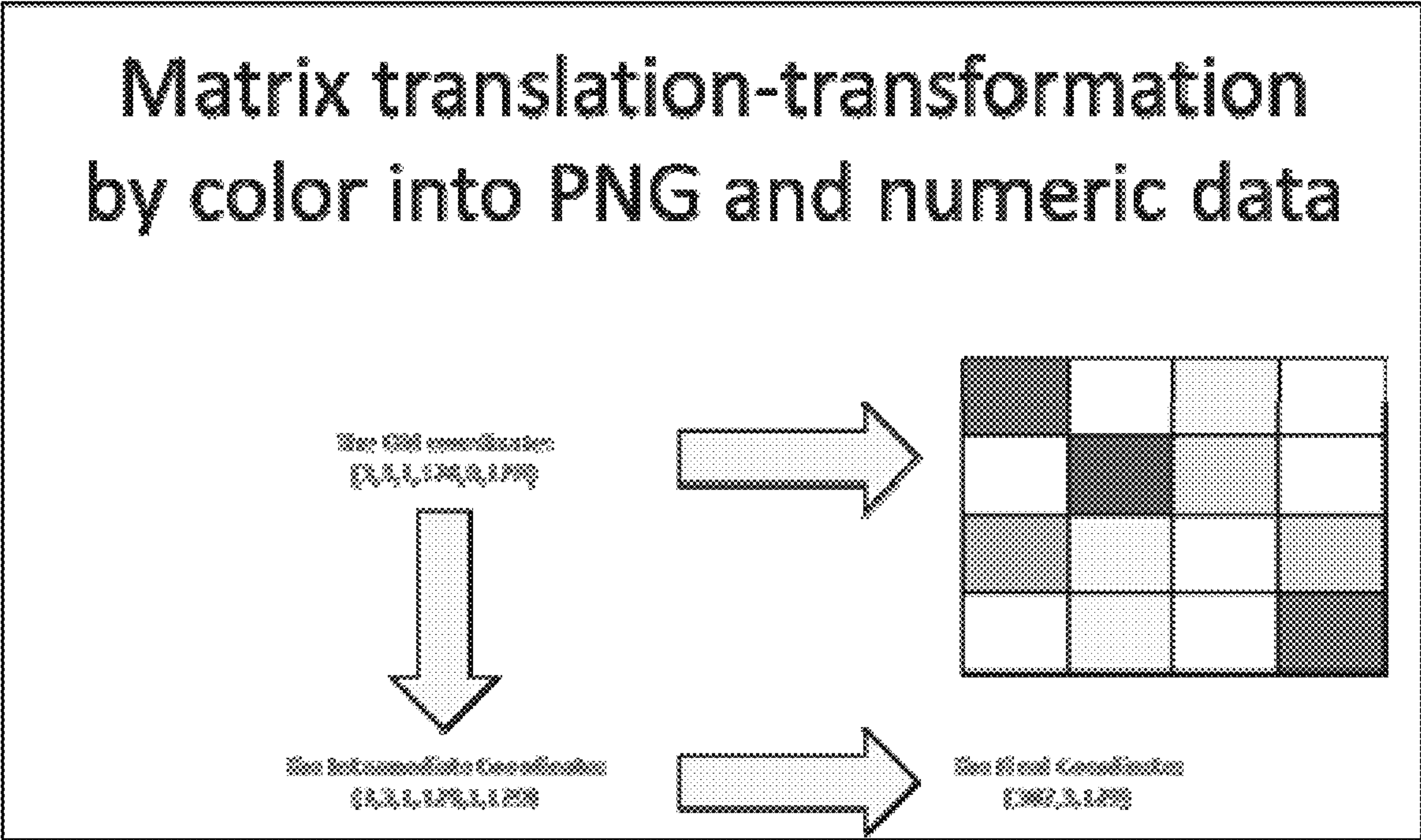


FIG. 14A

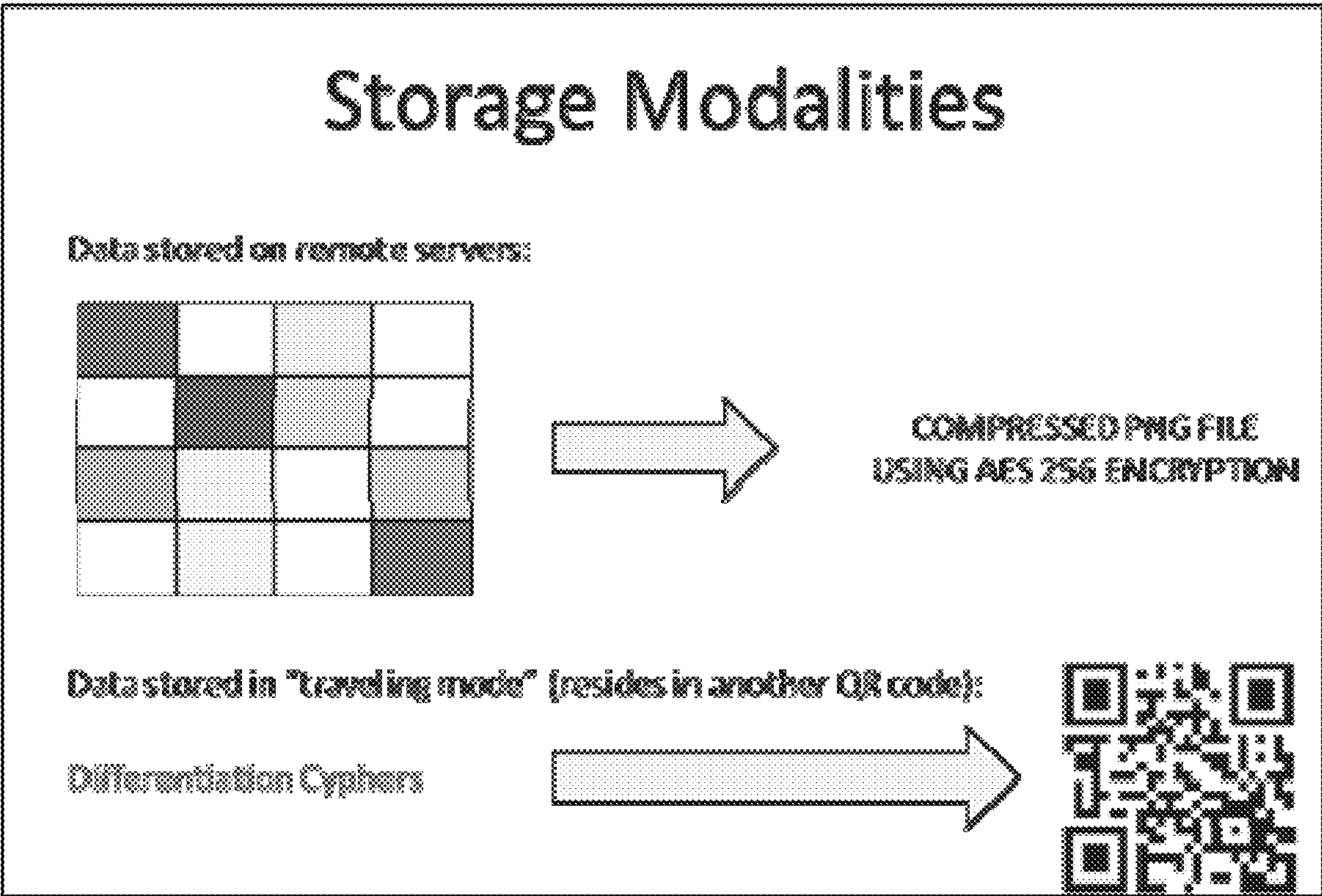
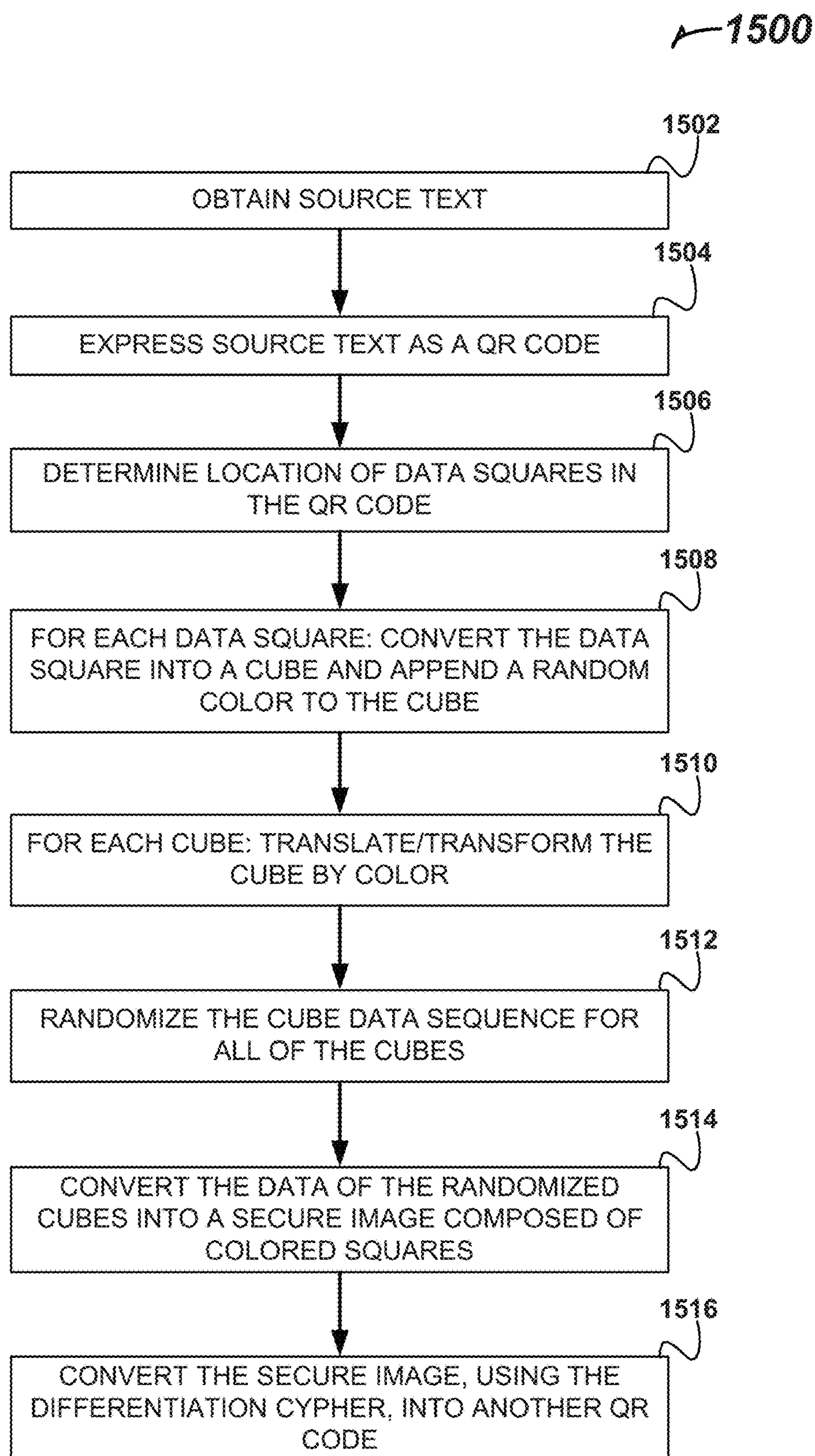


FIG. 14B

**FIG. 15**

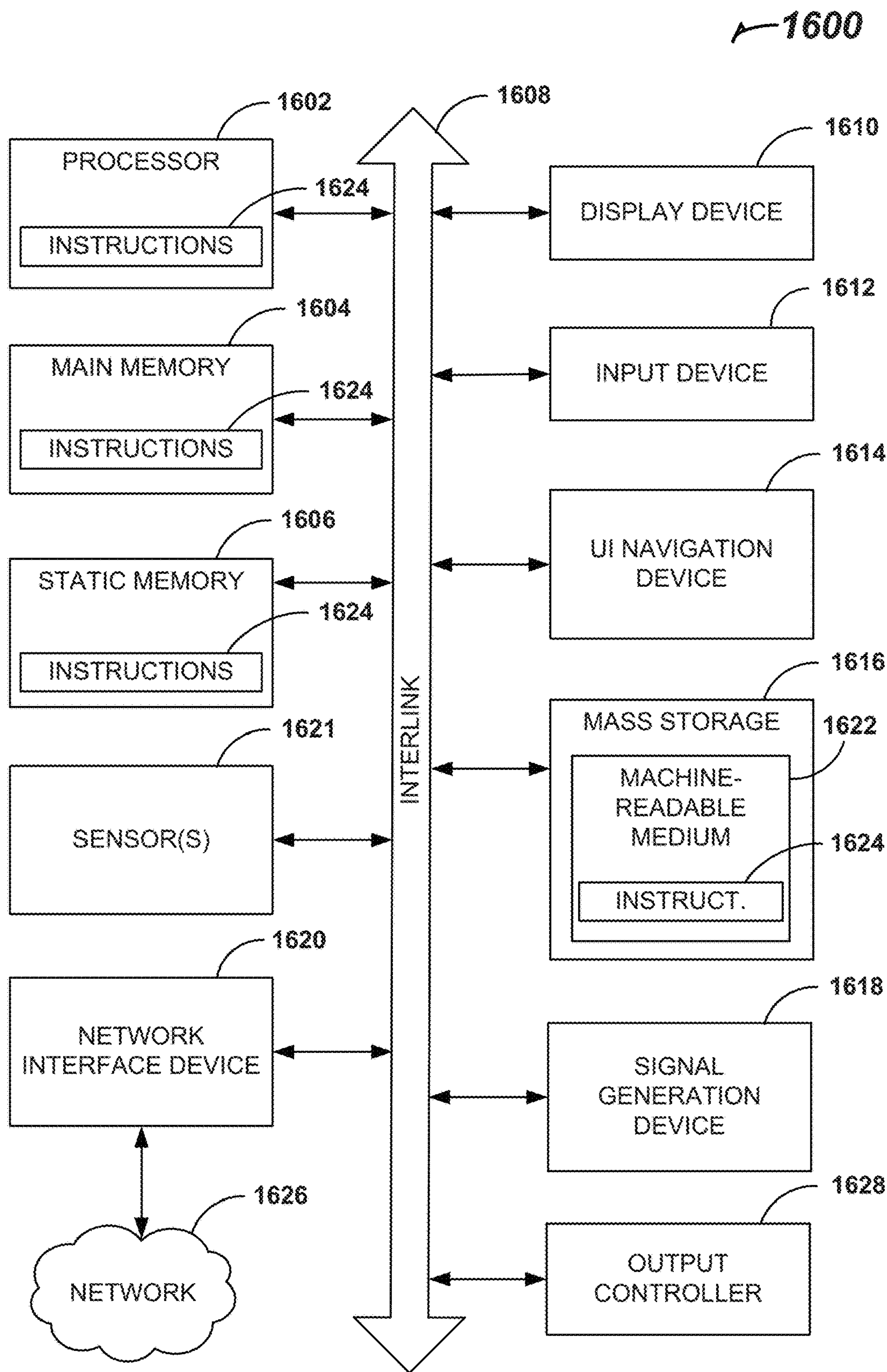


FIG. 16

AUTHENTICATION VIA TYPING CADENCE, GESTURES, AND QR CODES

CLAIM OF PRIORITY

This patent application is a Continuation of U.S. patent application Ser. No. 15/134,348, filed on Apr. 20, 2016, which claims the benefit of priority under 35 U.S.C. § 119(e) to U.S. Provisional Patent Application Ser. No. 62/150,213, filed on Apr. 20, 2015, each of which are hereby incorporated by reference herein in their entireties.

CROSS-REFERENCE TO RELATED PATENT DOCUMENTS

This patent application is also related to U.S. patent application Ser. No. 15/248,174, titled “User Authentication,” filed on Aug. 26, 2016; U.S. patent application Ser. No. 14/052,562, titled, “User Authentication via Input of Known Text,” filed on Oct. 11, 2013; and to U.S. Provisional Patent Application Ser. No. 61/712,718, titled, “System and Method to Differentiate Input Device Users,” filed on Oct. 11, 2012, each of which are hereby incorporated by reference herein in their entireties.

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BACKGROUND

As people’s work and personal lives become increasingly dependent upon computing devices and information networks, authenticating users who seek to access computing devices, computer networks, and computer services becomes increasingly important.

BRIEF DESCRIPTION OF THE DRAWINGS

Some embodiments are illustrated by way of example and not limitation in the various figures provided in-line with the text and in the figures of the accompanying drawings, in which:

FIG. 1A illustrates components of the Passphrase system, according to an embodiment;

FIG. 1B illustrates a user interface (UI) control that may be used to designate the previously collected cadence and habit library (PCCHL) text source(s) to be used and display an overview of data collected during a typing collection session, according to an embodiment

FIG. 2 illustrates a visual output including an example table that displays data of performance metrics created during the process of evaluating the challenge phrase against the user’s PCCHL, according to an embodiment;

FIG. 3 illustrates a visual output including a graph of hallmarks, according to an embodiment.

FIG. 4 illustrates a visual output including an alignment graph displaying overlap between a challenge phrase and a PCCHL, according to an embodiment;

FIG. 5 illustrates a visual output including a composite graphic of graphical elements which, together, visually depict the nature of the match between a challenge phrase sample set and an associated PCCHL, according to an embodiment;

FIG. 6 illustrates a visual output including two respective graphs for the Empirical Mode Decomposition (“EMD”) of the challenge phrase and the associated PCCHL, according to an embodiment;

FIG. 7A illustrates a visual output displaying additional statistics produced by the Passphrase system, according to an embodiment;

FIG. 7B is a flowchart of a method for user authentication via the Passphrase system, according to an embodiment;

FIG. 8 illustrates some of the metrics collected during a gesture authentication attempt, according to an embodiment;

FIG. 9 illustrates converting metrics collected during the gesture authentication attempt into an “effort straw,” according to an embodiment;

FIG. 10 illustrates a gesture composition during an authentication enrollment of a user, according to an embodiment;

FIG. 11 illustrates authentication gestures evaluated by an effort straw, according to an embodiment;

FIG. 12 illustrates expressing raw data as a QR code, according to an embodiment;

FIG. 13 illustrates converting a data box of a QR code into six dimensions, according to an embodiment;

FIG. 14A illustrates converting a cube represented by six dimensions into a “final coordinate” of three dimensions, according to an embodiment;

FIG. 14B illustrates two storage modalities of the disclosed QR data representation, according to an embodiment;

FIG. 15 is a flowchart of a method for obfuscating and storing data in a QR code, in accordance with some embodiments; and

FIG. 16 is a block diagram illustrating a machine in the example form of a computer system, within which a set or sequence of instructions for causing the machine to perform any one of the methodologies discussed herein may be executed, in accordance with some embodiments.

DETAILED DESCRIPTION

The present disclosure describes techniques for user authentication via typing cadence, gestures, and QR codes. In the following detailed description, for purposes of explanation, numerous specific details are set forth in order to provide a thorough understanding of the various aspects of the presently disclosed subject matter. However, it will be evident to those skilled in the art that the presently disclosed subject matter may be practiced without these specific details. In other instances, well-known methods, procedures, and components have not been described in detail so as not to obscure the presently disclosed subject matter.

Passphrase

Classic passwords (e.g., passwords that a user must remember in order to be authenticated or to be granted access to a secured resource) may be discovered by others, may be inferred from widespread/repeated use in quasi-public social environments (e.g., social media sites), or may otherwise be obtained through hacking or some other illegal process. Password restriction is the practice of enforcing rules regarding password length and the use of special

characters and punctuation marks in passwords. User frustration increases with password complexity, but less so with password length. Although well intended, complicated passwords and elaborate password expiration schedules do not increase security. A password consisting of a string of meaningful words is called a “passphrase.” Passphrases reduce the weaknesses of classic passwords by making the string harder to crack (e.g., “guess”) and by reducing user frustration with complex passwords.

Despite proclamations of the death of passwords, strong resistance remains to abandoning passwords altogether. A better solution to dealing with the weaknesses of classic passwords is by not requiring a user to remember the user’s password, but instead to require the user to enter a usefully random password that is dynamically generated at the time of use (access), and which is then discarded. However, how can a user type such a “password” without knowing, a priori, what to type?

Passphrase is software that learns the user’s finger movements over a keyboard during typing efforts and stores this information into a previously collected cadence and habit library (“PCCHL”, as defined in U.S. patent application Ser. No. 14/052,562, titled, “User Authentication via Input of Known Text,” filed on Oct. 11, 2013) associated with the user. The efforts within the PCCHL are analyzed for patterns, and muscle movement patterns of the user that are either predictably consistent or predictably inconsistent are identified. Passphrase then generates English (or any other language) phrases that feature one or more of the user’s hallmarks, which are groups of characters of varying lengths that the user types accurately, consistently, and frequently, and which are reliably descriptive of the way that particular user types. Using a sentence construction engine and the generated phrases featuring the user’s hallmarks, Passphrase creates a one-time authentication challenge (e.g., a short phrase or sentence) comprising a set of one or more generated phrases featuring the user’s hallmarks. Passphrase prompts the user to type the one-time authentication challenge text and then compares the efforts, which the user expended in typing the one-time authentication challenge, to correlated effort metrics in the PCCHL associated with the user to determine whether the two sets of efforts statistically match.

By examining only the metrics of a user’s typing efforts, which are stored in the user’s PCCHL, Passphrase eliminates the requirements and weaknesses of classic passwords (e.g., a user having to remember a particular password). Because a generated passphrase expires after a single authentication, the generated passphrase is of no value in subsequent authentications, and thus does not need to be remembered nor kept secret. Furthermore, a generated passphrase may be set to expire as a function of time, thus further securing the generated passphrase.

The entries in a user’s PCCHL may accumulate over time and reflect a capture process, which focuses on a consistent style of typing (e.g., the user’s typing when using a particular program or particular type of program, such as Microsoft’s Word and Outlook vs. Excel or some form-based program). Passphrase may access the user’s PCCHL, and may compare the typing metrics in the one-time authentication challenge to corresponding typing metrics in the PCCHL, and may return a result (e.g., renders an opinion or a value) as to their similarity. Supplemental information in the result may indicate a quality of the statistics, so that a host device may take appropriate action based on at least one of the host device’s risk tolerance, the sensitivity of the app, data, or process that is being protected, or other business rule

considerations. In some example embodiments, after only a short period of deployment, the PCCHL will have sufficient usable data, thereby enabling Passphrase to generate a nearly limitless variety of confected sentences.

Some host devices may episodically use this one-time authentication challenge capability in connection with a specific request to secure data, a process, or an app. The host device may require a user to authenticate via the one-time authentication challenge, which may be set to appear at any time in any program, before launching a new app, or before exposing confidential data. Passphrase may communicate over web services and may present standard component-style interfaces for integration into existing software.

A user need not formally enroll into Passphrase. In fact, a user need not change the user’s normal program/app usage. Passphrase inconspicuously tracks the user’s typing behavior and stores the relevant metrics in the user’s PCCHL.

FIG. 1A illustrates components of the Passphrase system **102**, according to an embodiment. Passphrase **102** may be comprised of eight components: (1) a typing collector **104**, (2) a hallmark chooser **106**, (3) a quirk finder **108**, (4) a sufficiency analyzer **110**, (5) a passphrase generator **112**, (6) a challenger **114**, (7) an evaluator **116**, and (8) a reporter **118**.

(1) Typing Collector

The Typing Collector obtains metrics about the physical activity involved in a user touching keys on a keyboard. In an embodiment, the Typing Collector does not capture any content (e.g., Typing Collector stores no source text), thus, Typing Collector does not create privacy concerns. The Typing Collector may run as a standalone program, which may be activated whenever a user’s PCCHL needs to be created or modified. A PCCHL is unique to a particular user because a PCCHL contains captured identifying behavior patterns, which are unique to each person. The Typing Collector may run as an installed application, or it may run as a web application. In general, latency from a slow Internet connection does not cause difficulties for Passphrase. The metrics that Passphrase uses may be captured by the Typing Collector. These metrics may include a) 100 ns timestamps associated with the KeyPress, KeyDown, and KeyUp events, b) a hashed representation of various groups of characters in text (e.g., the “hallmarks”) that compose the words, and c) selected text or typing activity surrounding the words that the user typed. The Typing Collector may also note the position of each hallmark within words and sentences, along with overall information about the speed at which each hallmark was typed. The Typing Collector may also collect intra-sample timing “overlaps,” which occur during the actuation cycles of temporally adjacent keystrokes, and the status of the “PNACS keys” (the Caps Lock key, the Num Lock key, the Alt key, the Control key, and the Shift key).

(2) Hallmark Chooser

Theoretically, a large number of potential hallmarks may be associated with every typing sample. A hallmark may be an n-length string of characters selected because of the distinctive behavioral effort involved in typing the hallmark. Hallmarks may overlap each other.

The entire collection of hallmarks represents a rich source of data for mining users’ personal typing patterns. For example, the five character string “ABCDE” (unrealistically short, of course, but used here only as an example) may have one or more of these 15 candidate hallmarks: A, B, C, D, E, AB, BC, CD, DE, ABC, BCD, CDE, ABCD, BCDE, and ABCDE itself. When data from PNACS keys’ movements as well as preceding and trailing characters is included, the number of potentially usable hallmarks may become quite

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large. The Hallmark Chooser may use one or more of several filtering algorithms to select reliably coherent, as well as reliably incoherent, strings—those strings, which occur with sufficient frequency that a reasonable number of them may be fairly anticipated to occur in a normally-typed sample of a user's totality of typing, assuming that a reasonable amount of text exists in the total population of typed characters. The ratio of the number of reliable hallmark samples compared to the total character count in the entire population of typing may form the primary basis for the designation of a character group as being a statistically usable hallmark.

The Hallmark Chooser may use a sentence construction engine, which may have a vocabulary of over 800,000 nouns, verbs, adjectives, and adverbs, as well as selected other parts of speech, to choose hallmarks. The Hallmark Choose may also have zero or more of the following features:

- anonymous personalization
- counterculture to current trend to randomize passwords and make them complicated
- uses logic to select what is ideal in Passphrase to validate
- chooses phrases from rosters of popular sayings
- varies the length of the word and the number of repetitions to hone the answer
- accidental key repeat auto-press
- computes implied finger travel from timings of long-run pairs
- timings of hyphens and semi-colons included
- keyboard travel speed to see what is impossible/improbable for single-finger typists
- includes length of key hold
- includes timings for overlap/shift—tied into finger travel propensity to use Caps Lock for long-capped strings
- space surrounding timings
- space timings

(3) Quirk Finder

A quirk is a spurt of typing that a user performs consistently, frequently, and that has resulted in errors or were executed at significantly varying speeds.

The Quirk Finder searches text looking for elusive quirks—mistake strings which are frequent and significant enough to merit being cataloged as idiosyncratic. In contrast to the Hallmark Chooser, which strives to find as many hallmarks as possible, the Quirk Finder is much more selective because a normal user typing normally is believed to intend accuracy, whereas quirks represent finger movement failures, which a normal user is believed to intend to minimize. Hence, the probability of the occurrence of quirks is relatively lower than for hallmarks. Furthermore, the probability of encountering a quirk during the typing of a short-burst challenge phrase is low. Since Passphrase is concerned with distinguishing the cadence of typing efforts (rather than identifying the letters) in text patterns, should quirks occur in the overall user-typed character population, and also occur during the creation of the challenge phrase, that fact would be particularly significant.

(4) Sufficiency Analyzer

The purpose of the Sufficiency Analyzer is to calibrate the adequacy of the source text composing the PCCHL for use in reporting the statistical probability of a match between the PCCHL text and the text that the user types in the challenge phrase.

The Sufficiency Analyzer addresses two foundational requirements concerning the statistics of probability: (1) truly random sampling, and (2) having enough data.

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The Sufficiency Analyzer also investigates the following potential problem areas in the PCCHL that might affect the usability of the results:

Failed collection plan and/or collection circumstances.

Differing nature of the designated PCCHL text. Ideally, the same source application type (e.g., Microsoft Outlook) would be designated for the PCCHL text, although applications that have the same basic typing behavior (e.g., Microsoft Outlook and Microsoft Word) will also work.

A full range of characters is preferred (unlike Microsoft Excel, which normally has mostly numbers).

Special characters are ignored (e.g., function keys, numeric keypad, etc.)

Measuring text metrics entropy. An infinite amount of typing over an infinite amount of time produces unremarkable (e.g., non-user-defining) data. Not enough typing causes the statistics to fail, whereas too much typing causes the data to fail. The Sufficiency Analyzer searches for this “Goldilocks Zone.”

(5) Passphrase Generator

Hallmarks are more often letter groups than complete words. A challenge phrase is composed of an easy-to-type noun-verb-object text string, which makes it easy for the user to type the challenge in a natural way (e.g., without having to pause to remember a complicated series of words, or to have to memorize complex strings of characters which are not words and thus require the effort of thinking in addition to the reflexive, mostly involuntary effort of habitual typing.) Passphrase's typical operating mode is to measure keyboard effort, not the identity (e.g., accuracy) of characters typed. Using the sentence construction engine, Passphrase randomly selects whole words that feature a user's hallmarks, deploying them in a corresponding sequential position according to metrics of the user's PCCHL. For example, if a hallmark was “gh” and located in a word-ending position (i.e., at the end of a word), then words like “cough,” “enough,” “breakthrough,” “thorough,” “laugh,” etc., words with similar endings, or words that contain a positional variation (e.g., “ghost”), would be candidates for a challenge phrase. The Passphrase generator may propound a variety of short, everyday-like sentences using hallmarks, unknown to the user, to sift effort metrics. The Passphrase generator may also generate non-hallmark-bearing phrases, for example when multiple challenges are warranted.

(6) Challenger

The Challenger requests everyday-style sentences from the Passphrase generator. The user is presented (e.g., visually on a display, aurally by playing the sentence aloud through a text-to-speech system, etc.) with the random sentence(s) and asked to type the presented random sentence (s). The Challenger may have a set of “guardrail” rules, so that the Challenger has the ability to reject typing efforts before they are submitted to the Evaluator. The Challenger records both the hallmark and quirk metrics of the user's typing efforts and passes them to the Evaluator for statistical evaluation.

(7) Evaluator

The Evaluator compares the metrics returned by the Challenger to the associated metrics in the PCCHL, attempting to match hallmarks using various statistical techniques.

In an embodiment, both the challenge phrase and the associated PCCHL metrics must each successively meet these statistical tests: (the italicized adjectives below are subject to tuning parameters):

There must be enough source characters in each sample. Since the Passphrase Generator will create the challenge phrase, this will normally be the case.

Intra-sample hallmark consistency is required. Less than 20% of all hallmark candidates will be used (although this threshold is tunable by the host app). The availability of quirks is a function of the user's typing—no typing mistakes equals no quirks. Catching a matching quirk is a significant statistical plus.

There must be enough qualifying hallmarks in each sample. By definition, the challenge phrase has hallmarks; also by definition, the PCCHL has those same hallmarks (this is required to create the challenge phrase in the first place). However, the hallmarks must be typed with an effort close enough to the expected pattern to qualify.

Qualifying hallmarks must represent a sufficient percentage of the overall typing in each sample.

There must be enough overlapping qualifying hallmarks between the sample phrase and the PCCHL.

There must be sufficient combined hallmarks as a percentage of overall challenge size.

(8) Reporter

The Reporter assesses the results returned by the Evaluator. The Reporter may be programmed to: (1) reject results conclusively, (2) require another challenge phrase effort, possibly with a longer challenge phrase; or (3) start the process over again. The results are returned with numerical and/or graphical information to assist the host application in determining a probability that the typist who typed the challenge phrase is the same person who created the host-selected PCCHL.

FIG. 1B illustrates a user interface (UI) control **152** that may be used to designate the PCCHL text source(s) to be used and display an overview of data collected during a typing collection session, according to an embodiment. The tree control **154** on the left may be used to select a user/machine combination, which causes a list of Checkboxes to be displayed on the right. Each Checkbox represents an application that has executed on the user/machine combination while the Typing Collector was active.

FIG. 2 illustrates a visual output **200** including an example table **202** that displays data of performance metrics created during the process of evaluating the challenge phrase against the user's PCCHL, according to an embodiment. Many of these data, particularly the Empirical Mode Decomposition (EMD) percentages, are single-number indicators of a likely match.

FIG. 3 illustrates a visual output **300** including a graph **302** of hallmarks, according to an embodiment. The graph **302** displays the behavior patterns across the most significant hallmarks encountered in the challenge phrase sample and the PCCHL as lines **304**, **306**, respectively. Although lines **304**, **306** will almost never be identical, even a brief glance at these two lines **304**, **306** suggests that they reflect typing efforts of the same person.

FIG. 4 illustrates a visual output **400** including an alignment graph **400** displaying overlap between a challenge phrase and a PCCHL, according to an embodiment. Circle **402** reflects the total number of candidate hallmarks, circle **404** reflects the number of hallmarks in the challenge phrase sample, and circle **406** reflects the number of hallmarks in the associated PCCHL. Statistical data may be displayed, such as the percent of the total hallmarks each circle **404**, **406** composes and the degree of overlap between circles **404**, **406**.

FIG. 5 illustrates a visual output **500** including a composite graphic **501** of graphical elements which, together,

visually depict the nature of the match between a challenge phrase sample set and an associated PCCHL, according to an embodiment. Included in the graphic **501** is a first circle **502**, a second circle **504**, an overlap zone corresponding to the overlap between the first circle **502** and the second circle **504**, an X, and a bounding box **510**, which may change in size and/or sheer. There are also two percentages, which depict the results of a series of complicated statistical computations into numbers that are referred to as "fidelity" and "confidence," which are terms that do not correspond to formal definitions in the world of statistics. "Fidelity" and "confidence" should be viewed together. Values for "fidelity" and "confidence" at either end of the 0-100 spectrum reflect reliability, whereas values in the 40-60 range reflect uncertainty.

FIG. 6 illustrates a visual output **600** including two respective graphs **602**, **604** for the Empirical Mode Decomposition ("EMD") of the challenge phrase and the associated PCCHL, according to an embodiment. The two graphs **602**, **604** may be compared visually by using certain statistical smoothing techniques on the challenge phrase data set and the associated PCCHL efforts. These graphs **602**, **604** produce "correlation" numbers that are single-number probabilities that the typist who typed the challenge phrase is the same person who created the host-selected PCCHL.

FIG. 7A illustrates a visual output **700** displaying additional statistics produced by the Passphrase system **102**, according to an embodiment. Passphrase **102** may produce statistics that describe many behavioral traits exhibited during the typing of the challenge phrase and the associated PCCHL. For example, FIG. 7 illustrates the display of the overall typing speed and the time of day (on a 24-hour clock), etc.

FIG. 7B is a flowchart of a method **701** for user authentication via the Passphrase system, according to an embodiment.

Metrics corresponding to the typing habits of a user are obtained (operation **702**).

A set of hallmarks of the user's typing habits are chosen (operation **704**).

Optionally, a set of quirks in the user's typing habits are searched for and found (operation **706**).

The adequacy of source text (e.g., PCCHL) of the user's typing habits is calibrated (operation **708**).

A passphrase is generated incorporating at least one hallmark of the set of hallmarks (operation **710**). Optionally, the generated passphrase includes at least one quirk.

The generated passphrase as an authentication challenge (operation **712**).

Metrics collected during the typing of the challenge passphrase are evaluated for statistical similarity to the metrics of the user's typing habits (operation **714**).

Output regarding the statistical similarity of the challenge passphrase metrics to the user's metrics is provided (operation **716**).

Gestures

A growing percentage of the world's commerce is conducted via mobile technology, particularly smartphones. These transactions need to be protected just as much as traditional e-commerce transactions using non-mobile technology. A common approach for securing e-commerce transactions has been to use software-generated, one-time use passcodes. However, one-time use passcodes do not provide true authentication of a person, but rather simply verify a person's possession of a token. Furthermore, automatically generated tokens may be easily intercepted and relayed, for example by malware or man-in-the-middle attacks.

A number of new approaches for securing mobile e-commerce transactions currently exist or are in late-stage development in the market, including (a) simulating numeric keypads with advanced touch-detection electronics, (b) one-time use passcodes sent by text messaging (e.g., SMS), (c) advanced screen components and processing services that may accurately read fingerprints (e.g., “fingerprint scanners”), and (d) drawing pattern recognition. These approaches, however, each have their drawbacks. For example, a drawn pattern may be seen by other people as it is being drawn, either in person or via image capturing equipment (e.g. video cameras), and may be described to others by word-of-mouth. If a drawing pattern is so complicated that it cannot be easily observed or described, then it is likely that the user attempting authentication will have difficulty recreating this drawing pattern dependably, and thus will encounter failure to acquire (“FTA”) problems. Conversely, if the pattern is simple, the pattern will be easy for others to replicate, and thus insecure.

The mountains of data—often exceeding 10,000 points from even sketching a simple circle—involved in just one finger movement of an authentication gesture may be successfully processed using the disclosed geospatial/Barycentric techniques that have previously been applied to keyboard typing. The effort of making the drawing is quantified, rather than the visual results/appearance of the drawing. Just as a potential impostor may observe a user typing the user’s passcode but cannot physically type the passcode with the same typing cadence as the user (thus rendering moot the potential impostor’s knowledge of the user’s passcode), a potential impostor may observe a user’s gesture, but cannot physically perform the gesture as the user does (thus rendering moot the potential impostor’s knowledge of the user’s gesture). Similar to using typing cadence versus typed characters for authentication, authentication using gesture efforts versus gesture sketches is a superior approach.

In an embodiment, the gesture authentication capability may be a separate application/app executing on a host device, or may be integrated into other applications/apps executing on the host device. Gesture authentication may be divided into three steps: (1) capturing the gesture metrics; (2) evaluating an instance (e.g., “sketch”) for a match against a PCCHL; and (3) reporting statistics of the evaluation to the host device. Deployments of gesture authentication generally fall into one of three scenarios: (1) all components executing on the host device; (2) capturing gesture metrics on the host device and the evaluation process executing remotely, with data exchange occurring via web services; and (3) capturing gesture metrics on the host device, and the evaluation process executing either locally on a PC or remotely on a server.

The enrollment process (e.g., capturing the base metrics for the PCCHL) involves the user first drawing a sample sketch of the sketch the user would like to use for future authentications. As with typing, this sample sketch should reflect comfortable movement by the user and should be simple enough to be remembered by the user. Some users may want to use long/complicated gestures, however, such gestures are not likely to be successful in everyday use; thus, a set of complexity limits may be imposed upon the enrollment process. As with typing, the simpler a gesture is, the better. After the sample sketch has been accepted, the user creates a PCCHL of substantially similar sketches. Each substantially similar sketch in the PCCHL serves as its own “White Hat” test, ensuring that the user can draw the sketch repeatedly and easily by making an acceptably similar effort.

This gesture authentication technique is unique in at least two ways. First, this gesture authentication technique effectively ignores the shape and appearance of the drawing, and instead looks to the effort of creating a similar drawing. Second, this gesture authentication technique processes metrics using a concept found in quantum physics: the graviton. A host device (e.g., a smartphone) during the course of one gesturing effort generates so many numbers across so many different degrees of freedom that the numbers must be processed as assemblages of clusters. Clusters, in turn, are composed of weighted rollups of smaller metrics groups—gravitons. Each “graviton” is composed of the combination of elements t_0 -X-Y-Z-R-P-Y-La-Lo-Al-V-F- t_n (e.g., start time, X-Axis, Y-Axis, Z-axis, Roll, Pitch, Yaw, Latitude, Longitude, Altitude, Velocity, Force, end time). The “glue” that holds these elements together is evidence of how each element was introduced into the sketch—the process and mechanics of the drawing effort. The proximity of these graviton clusters, coupled with the density of their mass, can be thought of as creating a form of gravity. The stronger the gravity is, the stronger the attraction is . . . and, thus, the stronger the affinity of the effort that created those clusters. Thus, sufficient affinity equals authentication.

This gesture authentication technique offers the benefits of a gradient result, bookended by absolutes, thus providing the host device with the option to permit n number of retries based on how close the drawing effort came to the expected metrics in the PCCHL.

FIG. 8 illustrates some of the metrics collected during a gesture authentication attempt, according to an embodiment. During the authentication sketch, the gesture authentication technique collects metrics produced by one or more devices integrated within the host device. Examples of such devices include accelerometers, gyroscopes, compasses (e.g., magnetometers), etc. The metrics produced by these devices and collected during the authentication sketch may be used in the evaluation/comparison of the authentication sketch against the PCCHL, and may be determinative of the host’s decision whether to authenticate the user drawing the sketch. For example, user A’s PCCHL includes metrics that indicate user A holds a smartphone at approximately 350 during user A’s authentication sketches. If a user attempting to authenticate as user A did not hold the smartphone at approximately 35° during the user’s authentication sketch, the gesture authentication technique may deny access to the user despite the user successfully reproducing user A’s authentication sketch.

FIG. 9 illustrates converting metrics collected during the gesture authentication attempt into an “effort straw,” according to an embodiment.

FIG. 10 illustrates a gesture composition during an authentication enrollment of a user, according to an embodiment. The drawn gesture **1002** is the gesture a user entered as the user’s unique gesture during enrollment into the gesture authentication system. After entering the drawn gesture **1002**, the system calculates the “effort straws” **1004** that correspond to the drawn gesture **1002**.

FIG. 11 illustrates authentication gestures evaluated by an effort straw **1102**, according to an embodiment. When a user attempts authentication via an authentication gesture, the gesture authentication system evaluates the authentication gesture through the effort straw **1102** previously calculated and stored in the user’s PCCHL. The geospatial/Barycentric algorithms evaluate the data corresponding to the authentication attempt and determine whether the authentication attempt was sufficiently within the effort straw **1102** stored in the user’s PCCHL. As illustrated in FIG. 11, an authen-

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tication attempt via gesture may be considered an “arrow” being shot through the effort straw **1102**. Arrow **1110** represents an authentication attempt whose gesture perfectly matches the authentication gesture in the user’s PCCHL. In actual conditions, a perfect match is nearly impossible. Arrow **1112** represents an authentication attempt whose gesture is sufficiently within the effort straw **1102** to be considered a match to the authentication gesture in the user’s PCCHL; this authentication would pass/succeed. Finally, arrow **1114** represents an authentication attempt whose gesture is NOT sufficiently within the effort straw **1102** to be considered a match to the authentication gesture in the user’s PCCHL; this authentication would fail.

This gesture authentication technique is device independent, in that it may be used by any number of devices. The authentication sketches may be scalable. For example, an authentication sketch performed on a 5-inch touchscreen display may be proportionally drawn on a 10-inch touchscreen display. In an embodiment, the host device uses a 1 ms or faster clock and has a touchscreen display that supports multi-touch input.

QR Codes

A QR code may be used to store user information, such as a user’s PCCHL, thus enabling a host device to authenticate a user without the host having to perform network communications. This may be useful in various situations, such as when Internet connectivity to the host device is slow or unavailable, or where a fast authentication check is required. This also allows a PCCHL to be physically portable.

A QR code may be used to recreate another QR code. The error correction in ISO standard (ISO/IEC18004) for QR codes allows a QR code to have an error rate of up to 30% without a loss of data. Thus, the squares of a QR code may be “selectively damaged” to encode data, for example facts and/or questions relating to a user’s PCCHL. A standard QR code may be configured, transmitted, and deployed in a way that stores information without other systems or people being able to decipher the QR code’s content. Therefore, the QR code may function as a “server,” enabling rapid authentication to protect confidential data, classified apps, restricted processes, or for other authentication purposes.

Although there are a large number of scenarios for which QR codes are ideally suited, one particularly suited application of QR codes is in combination with gesture-based authentication. Summary gesture metrics require a relatively small number of bytes, which fit inside the size limitations of a QR code. When a user draws a sketch during an authentication, the effort to draw the sketch is compared to the user’s PCCHL, which contains similar efforts of that user. A properly formatted QR code may be used to store the user’s PCCHL.

The QR authentication technology enables the host device (e.g., a computer protected by the previously described gesture authentication technology) to “request” a user’s gesture simply by displaying (e.g., via an application executing on the host device) a properly formatted QR code. When the user sees the QR code, the user may scan the QR code with the user’s smartphone app, which then invites the user to draw the expected sketch on the smartphone. Although the app will accept the sketch input, in an embodiment, the app will not display the sketch on the app’s user interface to prevent unintentional observation by potential third parties. At the conclusion of the drawing effort, the smartphone app may return a report on how the drawing effort compared to other efforts in the user’s PCCHL. The user may consume this report by one or more means, such as web services,

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Bluetooth/NFC (Near Field Communication), or directly in the app running on the host device.

Another potential use case is as follows: suppose the host device has a stored document that has been divided into sections. a respective section of the document has a minimum respective clearance level; a person with at least that respective clearance level is allowed to view the respective section of the document. A user selects the document for viewing. In response, the host device displays the QR code. The user scans the QR code with the user’s smartphone, then traces the user’s authentication gesture on the user’s smartphone. Based on the nature and metrics of the effort expended in the user’s response—and hence the user’s authenticated status—access is selectively granted to various parts of the document without the host device ever knowing what was drawn and without the user’s gesture being seen by anyone else.

App developers may embed this QR authentication capability into their own app workflow processes, presumably in a solicitation stage and a follow-on results-processing stage. The QR authentication technology may be available as deployable modules (e.g., dynamically linked libraries or “DLLs”) with industry standard interfaces, making them easy to integrate, deploy, and use.

FIG. 12 illustrates expressing raw data as a QR code, according to an embodiment. In the illustrated example, the string “ABC” (the raw data **1202**) is expressed as a QR code **1204** with nine shaded “data boxes.” Although the data boxes are shaded in FIG. 12, a data box in a typical QR code would be solid black. Each “data box” may be considered to be one “unit” square, and has an “address” corresponding to some coordinate system (e.g., Cartesian).

FIG. 13 illustrates converting a data box of the QR code **1204** into six dimensions, according to an embodiment. Each “data box” of the QR code **1204** is assigned a random color and is transformed into a cube, whose “location” is expressed in two sets of dimensions: one set representing the physical dimensions of the cube and another set of dimensions representing the color of the cube. For example, if a Cartesian coordinate system (e.g., X, Y, and Z) is being used to represent physical dimensions and a Red-Green-Blue color model (e.g., RGB) is being used to represent color, each cube would have six dimensions: (X,Y,Z,R,G,B).

For example, the top-left corner of the bottom-right-most data box in the QR code **1204** has Cartesian coordinates (3,1); adding a third-dimension to this data box results in the (X,Y,Z) coordinates (3,3,1) for the corresponding cube **1304**. The RGB values (128,0,128), which represent the color purple, were chosen at random and assigned to the cube **1304**. Thus, the cube **1304** corresponding to the bottom-right-most data box in the QR code **1204** in this example has (X,Y,Z,R,G,B) coordinates of (3,3,1,128,0,128).

FIG. 14A illustrates converting a cube represented by six dimensions into a “final coordinate” of three dimensions, according to an embodiment. For each data box of the QR code **1204**, the six dimensional value (X,Y,Z,R,G,B) of its corresponding cube is translated into its corresponding final coordinate by compressing it back into three dimensions (X',Y',Z'). In an embodiment, the compression may use the formula $X'=X*R$, $Y'=Y*G$, and $Z'=Z*B$. In an embodiment, the physical dimension coordinates and/or the color values are adjusted by adding 1, thus avoiding multiplication by zero. For example, the (X,Y,Z,R,G,B) coordinates (3,3,1,128,0,128) corresponding to data box (3,1) of the QR code **1204** would be converted as follows:

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1. The RGB values of (3,3,1,128,0,128) are incremented by 1, resulting in an intermediate coordinate value of (3,3,1,129,1,129).

2. The intermediate coordinate value of (3,3,1,129,1,129) is converted into a final coordinate value: $X'=X*R=3*129=387$; $Y'=Y*G=3*1=3$; and $Z'=Z*B=1*129=129$. Thus, the final coordinate value (X', Y', Z') is (387,3,129).

FIG. 14B illustrates two storage modalities of the disclosed QR data representation, according to an embodiment. The final coordinate for a data box of the QR code 1204 is ported to a Differentiation Cypher, which is created using the combination of: (1) the creation time of the source data (e.g., "ABC"), (2) the QR Code Version and Correction Level, and (3) the position of a given Differentiation Cypher in the final text string to map RGB color to a final coordinate. For example, the Differentiation Cyphers of all non-white data boxes of the QR code 1204 are converted into fixed-length text strings, whose text is then expressed as a QR code.

FIG. 15 is a flowchart of a method 1500 for obfuscating and storing data in a QR code, according to an embodiment.

A source text (e.g., a password) is obtained (operation 1502).

The source text is expressed in a QR code (operation 1504).

The location of the "data squares" is determined (operation 1506).

Each data square is converted into a cube, and a random color is appended to the cube (operation 1508).

Each cube is translated/transformed by color (operation 1510).

The cube data sequence for all of the cubes is randomized (operation 1512).

The data of the randomized cubes is saved to a server as a secure image composed of colored squares (operation 1514) and/or

Optionally, the secure image is converted, using the Differentiation Cypher as the color-to-location mapper, into another QR code (operation 1516).

Some Distinguishing Characteristics of the Disclosed QR Functionality:

- Data format change through QR
- Data confusion (lossless) in correction code
- Data format change through color
- Data sequence immaterial—Jackson Pollock style (random color splotches)
- Arbitrary color assignment by random Tick seed
- No reference to anything outside the QR code
- Identical source data always appears differently
- QR has Six Main Technology Aspects:

"Use" related:

- contain a passphrase
- describe the gesture, rather than create it
- propound cryptographic nonces—[list]
- read a printed QR code (e.g., carried in a wallet) to provide the user's PCCHL to another host device

"Send" related:

- Refer to the islands by either latitude/longitude or by position number (e.g., a left-to-right then top-to-bottom winding string)

Has the ability to chain QR codes

"Programming" related:

- can be written to run on a smartphone, a host application, or on a web services-enabled server

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"Limiting" related:

- can perform geofencing
- barcode painting can be hash-tagged to facilitate speedy identification of duplicate second attack
- duplicate rejection performed (variable time window)
- QR code lifecycle (availability) can be bounded by geofencing or elapsed time or usage count

"Deforming" related:

- alternate between painting white islands and black islands
- blobs painted in random order
- contains random noise
- creates executable code to run a hidden program—interpolated execution
- drive a speech engine to propound question by sound point to a hidden URL
- ability to deform can itself be randomized

"Color" related:

- color can be preserved in the QR bitmap
- level of color can be determined by normal cameras, like those on smartphones—does not require a 'professional' grade camera
- color is another dimension
- color can signal which characters to ignore

FIG. 16 is a block diagram illustrating an example of a machine 1600, upon which any example embodiments may be implemented. In an embodiment, the machine 1600 is connected (e.g., networked) to other machines. In a networked deployment, the machine 1600 may operate in the capacity of a server machine, a client machine, or both a client machine and a server machine in server-client network environments. In an embodiment, the machine 1600 acts as a peer machine in peer-to-peer (P2P) (or other distributed) network environment. In an embodiment, machine 1600 is a network router, switch, or bridge, or any machine capable of executing instructions (sequential or otherwise) that specify actions to be taken by that machine. Further, while only a single machine is illustrated, the term "machine" shall also be taken to include any collection of machines that individually or jointly execute a set (or multiple sets) of instructions to perform any one or more of the methodologies discussed herein, such as cloud computing, software as a service (SaaS), other computer cluster configurations.

Examples, as described herein, may include, or may operate by, logic or a number of components, or mechanisms. Circuit sets are a collection of circuits implemented in tangible entities that include hardware (e.g., simple circuits, gates, logic, etc.). Circuit set membership may be flexible over time and underlying hardware variability. Circuit sets include members that may perform, alone or in combination, specified operations when operating. In an example, hardware of the circuit set may be immutably designed to carry out a specific operation (e.g., hardwired). In an example, the hardware of the circuit set may include variably connected physical components (e.g., execution units, transistors, simple circuits, etc.) including a machine-readable medium physically modified (e.g., magnetically, electrically, moveable placement of invariant massed particles, etc.) to encode instructions of the specific operation. In connecting the physical components, the underlying electrical properties of a hardware constituent are changed, for example, from an insulator to a conductor or vice versa. The instructions enable embedded hardware (e.g., the execution units or a loading mechanism) to create members of the circuit set in hardware via the variable connections to carry out portions of the specific operation when in operation. Accordingly, the machine-readable medium is communicatively coupled to the other components of the circuit set member when the device is operating. In an example, any of

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the physical components may be used in more than one member of more than one circuit set. For example, under operation, execution units may be used in a first circuit of a first circuit set at one point in time and reused by a second circuit in the first circuit set, or by a third circuit in a second circuit set at a different time.

Machine (e.g., computer system) **1600** may include a hardware processor **1602** (e.g., a central processing unit (CPU), a graphics processing unit (GPU), a hardware processor core, or any combination thereof), a main memory **1604** and a static memory **1606**, some or all of which may communicate with each other via an interlink (e.g., bus) **1608**. The machine **1600** may further include a display device **1610**, an input device **1612** (e.g., buttons, switches, a keyboard, etc.), and a user interface (UI) navigation device **1614** (e.g., a pointing stick, an isometric joystick or other isometric device, etc.). In an embodiment, the display device **1610**, input device **1612**, and UI navigation device **1614** may be a touch screen display. The machine **1600** may additionally include a storage device (e.g., drive unit) **1616**, a signal generation device **1618** (e.g., a speaker), a network interface device **1620**, and a sensor **1621**, such as a global positioning system (GPS) sensor, compass, accelerometer, or other sensor. The machine **1600** may include an output controller **1628**, such as a serial (e.g., universal serial bus (USB), parallel, or other wired or wireless (e.g., infrared (IR), near field communication (NFC), etc.) connection to communicate or control one or more peripheral devices (e.g., a printer, card reader, etc.)

The storage device **1616** may include a machine-readable medium **1622**, on which is stored one or more sets of data structures or instructions **1624** (e.g., software) embodying or utilized by any one or more of the techniques or functions described herein. The instructions **1624** may also reside, completely or at least partially, within the main memory **1604**, within static memory **1606**, or within the hardware processor **1602** during execution thereof by the machine **1600**. In an example, one or any combination of the hardware processor **1602**, the main memory **1604**, the static memory **1606**, or the storage device **1616** may constitute machine-readable media.

Although the machine-readable medium **1622** is illustrated as a single medium, the term “machine-readable medium” may include a single medium or multiple media (e.g., a centralized or distributed database, and/or associated caches and servers) arranged to store the one or more instructions **1624**.

The term “machine-readable medium” may include any medium that is capable of storing, encoding, or carrying instructions for execution by the machine **1600** and that cause the machine **1600** to perform any one or more of the techniques of the present disclosure, or that is capable of storing, encoding or carrying data structures used by or associated with such instructions. Non-limiting machine-readable medium examples may include solid-state memories, and optical and magnetic media. In an example, a massed machine-readable medium comprises a machine-readable medium with a plurality of particles having invariant (e.g., rest) mass. Accordingly, massed machine-readable media are not transitory propagating signals. Specific examples of massed machine-readable media may include non-volatile memory, such as semiconductor memory devices (e.g., Electrically Programmable Read-Only Memory (EPROM), Electrically Erasable Programmable Read-Only Memory (EEPROM)) and flash memory

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devices; magnetic disks, such as internal hard disks and removable disks; magneto-optical disks; and CD-ROM and DVD-ROM disks.

The instructions **1624** may further be transmitted or received over a communications network **1626** using a transmission medium via the network interface device **1620** utilizing any one of a number of transfer protocols (e.g., frame relay, internet protocol (IP), transmission control protocol (TCP), user datagram protocol (UDP), hypertext transfer protocol (HTTP), etc.). Example communication networks may include a local area network (LAN), a wide area network (WAN), a packet data network (e.g., the Internet), mobile telephone networks (e.g., cellular networks), Plain Old Telephone (POTS) networks, and wireless data networks (e.g., Institute of Electrical and Electronics Engineers (IEEE) 802.11 family of standards known as Wi-Fi, IEEE 802.16 family of standards known as WiMax®, IEEE 802.15.4 family of standards, IEEE 802.15.1 family of standards (known as Bluetooth®), peer-to-peer (P2P) networks, among others. In an embodiment, the network interface device **1620** may include one or more physical jacks (e.g., Ethernet, coaxial, or phone jacks) and/or one or more antennas to connect to the communications network **1626**. In an example, the network interface device **1620** may include a plurality of antennas to wirelessly communicate using at least one of single-input multiple-output (SIMO), multiple-input multiple-output (MIMO), or multiple-input single-output (MISO) techniques. The term “transmission medium” shall be taken to include any intangible medium that is capable of storing, encoding, or carrying instructions for execution by the machine **1600**, and includes digital or analog communications signals or other intangible medium to facilitate communication of such software.

The various operations of the example methods described herein may be performed, at least partially, by one or more processors that are temporarily arranged (e.g., by software instructions) or permanently arranged to perform the relevant operations. Whether temporarily or permanently arranged, such processors may constitute processor-implemented modules or objects that operate to perform one or more operations or functions. The modules and objects referred to herein, in some example embodiments, may comprise processor-implemented modules and/or objects.

Similarly, the methods described herein may be at least partially processor-implemented. For example, at least some of the operations of a method may be performed by one or more processors or processor-implemented modules. The performance of certain operations may be distributed among the one or more processors, not only residing within a single machine or computer, but also deployed across a number of machines or computers. In some example embodiments, the processor or processors may be located in a single location (e.g., within a home environment, an office environment, at a server farm, etc.), while in other embodiments, the processors may be distributed across a number of locations.

The one or more processors may also operate to support performance of the relevant operations in a “cloud computing” environment or within the context of software as a service (SaaS). For example, at least some of the operations may be performed by a group of computers (as examples of machines including processors), these operations being accessible via a network (e.g., the Internet) and via one or more appropriate interfaces (e.g., Application Program Interfaces (APIs)).

Additional Notes

Conventional terms in the fields of computer networking and computer systems have been used herein. The terms are

known in the art and are provided only as a non-limiting example for convenience purposes. Accordingly, the interpretation of the corresponding terms in the claims, unless stated otherwise, is not limited to any particular definition.

Although specific embodiments have been illustrated and described herein, it will be appreciated by those of ordinary skill in the art that any arrangement that is calculated to achieve the same purpose may be substituted for the specific embodiments shown. Many adaptations will be apparent to those of ordinary skill in the art. Accordingly, this application is intended to cover any adaptations or variations.

The above detailed description includes references to the accompanying drawings, which form a part of the detailed description. The drawings show, by way of illustration, specific embodiments that may be practiced. These embodiments are also referred to herein as “examples.” Such examples may include elements in addition to those shown or described. However, the present inventors also contemplate examples in which only those elements shown or described are provided. Moreover, the present inventors also contemplate examples using any combination or permutation of those elements shown or described (or one or more aspects thereof), either with respect to a particular example (or one or more aspects thereof), or with respect to other examples (or one or more aspects thereof) shown or described herein.

All publications, patents, and patent documents referred to in this document are incorporated by reference herein in their entirety, as though individually incorporated by reference. In the event of inconsistent usages between this document and those documents so incorporated by reference, the usage in the incorporated reference(s) should be considered supplementary to that of this document; for irreconcilable inconsistencies, the usage in this document controls.

In this document, the terms “a” or “an” are used, as is common in patent documents, to include one or more than one, independent of any other instances or usages of “at least one” or “one or more.” In this document, the term “or” is used to refer to a nonexclusive or, such that “A or B” includes “A but not B,” “B but not A,” and “A and B,” unless otherwise indicated. Moreover, in the following claims, the terms “first,” “second,” and “third,” etc. are used merely as labels, and are not intended to impose numerical requirements on their objects.

Method examples described herein may be machine or computer-implemented at least in part. Some examples may include a machine-readable medium or machine-readable medium encoded with instructions operable to configure an electronic device to perform methods as described in the above examples. An implementation of such methods may include code, such as microcode, assembly language code, a higher-level language code, or the like. Such code may include machine-readable instructions for performing various methods. The code may form portions of computer program products. Further, in an example, the code may be tangibly stored on one or more volatile, non-transitory, or non-volatile tangible machine-readable media, such as during execution or at other times. Examples of these tangible machine-readable media may include, but are not limited to, hard disks, removable magnetic disks, removable optical disks (e.g., compact disks and digital video disks), magnetic cassettes, memory cards or sticks, random access memories (RAMs), read-only memories (ROMs), and the like.

The above description is intended to be illustrative, and not restrictive. For example, the above-described examples (or one or more aspects thereof) may be used in combination

with each other. Other embodiments may be used, such as by one of ordinary skill in the art upon reviewing the above description. The Abstract is provided to allow the reader to quickly ascertain the nature of the technical disclosure and is submitted with the understanding that it will not be used to interpret or limit the scope or meaning of the claims.

In this Detailed Description, various features may have been grouped together to streamline the disclosure. This should not be interpreted as intending that an unclaimed disclosed feature is essential to any claim. Rather, inventive subject matter may lie in less than all features of a particular disclosed embodiment.

The invention claimed is:

1. A user authentication system, comprising:
 - a cadence library stored on a non-transitory computer-readable medium, wherein the cadence library corresponds to a user and has data corresponding to the user including base metrics associated with gestures of the user;
 - one or more non-transitory memory storage devices; and
 - one or more processors configured to execute instructions stored in the one or more non-transitory memory storage devices, which instructions, when executed by the processor, cause the user authentication system to perform operations to:
 - record gesture data captured from a sensor device in response to the sensor device detecting gesture input by the user;
 - process a combination of weighted metrics derived from the gesture data recorded, the combination correlated to proximity of the weighted metrics coupled with a density associated with the weighted metrics;
 - compare the combination of weighted metrics to the data corresponding to the user within the cadence library.
2. The user authentication system of claim 1, wherein the operations include:
 - segmentation of a user's input on a touchscreen sensor by time and vector changes mapped on a n-dimensional space composed of screen coordinates of the input, time, x-axis, y-axis, z-axis, roll, pitch, yaw, latitude, longitude, altitude, velocity, and force;
 - collection of the segments into clusters of data representing weighted rollup metrics, using normalization statistical techniques; and
 - assembling the clusters into effort metrics that describe a complete gesture representing a user's input on the touchscreen sensor.
3. The user authentication system of claim 1, wherein the gesture data is captured from multiple sensor devices.
4. The user authentication system of claim 1, wherein the operations include executing an enrollment process to capture the base metrics to form the cadence library corresponding to the user.
5. The user authentication system of claim 1, wherein the operations include:
 - production of an indicator of similarity from the comparison of the combination of weighted metrics to the data corresponding to the user within the cadence library;
 - generation, based on the comparison and one or more criteria for the comparison, of an indication signal representative of a determination that the combination of weighted metrics is a match with a collection of complete gestures contained in the cadence and habit library corresponding to the user; and

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generation of an expression of results of the determination as a set of metrics and derived summaries for subsequent analysis by an external process.

6. The user authentication system of claim 5, wherein the production of an indicator of similarity uses geospatial/ Barycentric algorithms. 5

7. The user authentication system of claim 1, wherein the operations include storage of the cadence library in the form of a quick response (QR) matrix barcode.

8. The user authentication system of claim 7, wherein the operations include use of coordinate math and randomization to express a secure derivative of the cadence library in the QR matrix barcode. 10

9. The user authentication system of claim 7, wherein the operations include generation of QR matrix barcode as defined in one or more standards. 15

10. The user authentication system of claim 7, wherein the operations include formation of the QR matrix barcode by: reading bytes from the cadence library; and converting the bytes to the QR matrix barcode. 20

11. The user authentication system of claim 7, wherein the operations include:

reading a QR matrix barcode image stored on non-transitory computer-readable medium or from a printed surface; 25

eliminating duplicate values from data that fall within an error correction capacity of ISO standard ISO/IEC18004; and

creating encoded image data from QR matrix barcode image. 30

12. The user authentication system of claim 11, wherein multiple QR matrix barcode images are used to reconstruct the bytes corresponding to the cadence library.

13. A user authentication system, comprising:

a cadence library stored on a non-transitory computer-readable medium, wherein the cadence library corresponds to a user and has data corresponding to the user including base metrics associated with gestures of the user; 35

one or more non-transitory memory storage devices; and one or more processors configured to execute instructions stored in the one or more non-transitory memory storage devices, which instructions, when executed by the processor, cause the user authentication system to perform operations to: 40

record gesture data captured from a sensor device in response to the sensor device detecting gesture input by the user; 45

process a combination of weighted metrics derived from the gesture data recorded; 50

compare the combination of weighted metrics to the data corresponding to the user within the cadence library;

store the cadence library in the form of a quick response (QR) matrix barcode; 55

form the QR matrix barcode by:

reading bytes from the cadence library; and converting the bytes to the QR matrix barcode;

transcode location of each black square in the QR matrix barcode to a two-dimensional coordinate using the form x, y ; 60

add a third dimension to the coordinate forming a first three-dimensional coordinate to transform the square into a cube represented in a three-dimensional space using the form x, y, z ; 65

associate the three-dimensional coordinate with a randomly-generated color value, where the color is

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expressed as a combination of red, green, and blue values, each in a numeric range between 1 and 256, taking the form r, g, b ;

append three more dimensions to the cube with the three more dimensions corresponding to the red, green, and blue values of the associated color, forming a second three-dimensional coordinate using the form r, g, b ;

generate a new derivative three-dimensional coordinate, using the form x', y', z' by multiplying the values of the first three dimensions by the corresponding values of the second three dimensions, such that $x'=x*r$, $y'=y*g$, and $z'=z*b$;

combine all the derivative three-dimensional coordinates corresponding to each of the black squares found in the QR matrix barcode, in newly-created encoded image data composed of colored squares corresponding to the derivative three-dimensional coordinates; and

produce a new, valid QR matrix barcode image from the encoded image data.

14. The user authentication system of claim 13, wherein the operations include operations to introduce a subset of random duplicate values from the set of derivative three-dimensional coordinates which do not exceed an error correction capacity described in an associated standard.

15. The user authentication system of claim 13, wherein multiple QR matrix barcode images are encoded that represent component parts of the bytes from the cadence library.

16. A user authentication system, comprising:

a cadence library stored on a non-transitory computer-readable medium, wherein the cadence library corresponds to a user and has data corresponding to the user including base metrics associated with gestures of the user;

one or more non-transitory memory storage devices; and one or more processors configured to execute instructions stored in the one or more non-transitory memory storage devices, which instructions, when executed by the processor, cause the user authentication system to perform operations to:

record gesture data captured from a sensor device in response to the sensor device detecting gesture input by the user;

process a combination of weighted metrics derived from the gesture data recorded;

compare the combination of weighted metrics to the data corresponding to the user within the cadence library;

store the cadence library in the form of a quick response (QR) matrix barcode;

read a QR matrix barcode image stored on non-transitory computer-readable medium or from a printed surface;

eliminate duplicate values from data that fall within an error correction capacity; and

create encoded image data from QR matrix barcode image;

build an array of dimensional values from color and coordinate positions of each square element that is part of the encoded image data, with the dimensional values taking a form of x', y', z' ;

convert the color of each element in the array of dimensional values into a corresponding set of red, green, and blue constituent values to be appended as

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additional dimensions, taking the form r, g, b, such that each dimensional value is expressed as x, y, z, r, g, b;

derive a collection of new coordinate values from the array of dimensional values, taking a form of x, y, z, 5
by calculating a quotient for each coordinate position in the array of dimensional values, divided by corresponding red, green, and blue values of the same dimensional value, such that $x=x'/r$, $y=y'/g$, and $z=z'/b$; 10

construct a new QR matrix barcode image from the collection of new coordinate values, with each coordinate representing a single black square in the QR matrix barcode image; and

interpret the value of the QR matrix barcode image as 15
bytes of the cadence library.

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